

# Spatiotemporal variability and contribution of different aerosol types to the Aerosol Optical Depth over the Eastern Mediterranean

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## Abstract

This study characterizes the spatiotemporal variability and relative contribution of different types of aerosols to the Aerosol Optical Depth (AOD) over the Eastern Mediterranean as derived from MODIS Terra (3/2000-12/2012) and Aqua (7/2002-12/2012) satellite instruments. For this purpose, a  $0.1^\circ \times 0.1^\circ$  gridded MODIS dataset was compiled and

1 validated against sunphotometric observations from the AErosol RObotic NETwork  
2 (AERONET). The high spatial resolution and long temporal coverage of the dataset allows for  
3 the determination of local hot spots like megacities, medium sized cities, industrial zones and  
4 power plant complexes, seasonal variabilities, and decadal averages. The average AOD at 550  
5 nm ( $AOD_{550}$ ) for the entire region is  $\sim 0.22 \pm 0.19$  with maximum values in summer and  
6 seasonal variabilities that can be attributed to precipitation, photochemical production of  
7 secondary organic aerosols, transport of pollution and smoke from biomass burning in Central  
8 and Eastern Europe, and transport of dust from the Sahara Desert and the Middle East. The  
9 MODIS data were analyzed together with data from other satellite sensors, reanalysis projects  
10 and a chemistry-aerosol-transport model using an optimized algorithm tailored for the region  
11 and capable of estimating the contribution of different aerosol types to the total  $AOD_{550}$ . The  
12 spatial and temporal variability of anthropogenic, dust and fine mode natural aerosols over  
13 land and anthropogenic, dust and marine aerosols over the sea is examined. The relative  
14 contribution of the different aerosol types to the total  $AOD_{550}$  exhibits a low/high seasonal  
15 variability over land/sea areas, respectively. Overall, anthropogenic aerosols, dust and fine  
16 mode natural aerosols account for  $\sim 51\%$ ,  $\sim 34\%$  and  $\sim 15\%$  of the total  $AOD_{550}$  over land,  
17 while, anthropogenic aerosols, dust and marine aerosols account  $\sim 40\%$ ,  $\sim 34\%$  and  $\sim 26\%$   
18 of the total  $AOD_{550}$  over the sea, based on MODIS Terra and Aqua observations.

19

## 20 **1 Introduction**

21 For more than fifteen years, two MODIS (Moderate Resolution Imaging Spectroradiometer)  
22 satellite sensors monitor tropospheric aerosols at a global scale on a daily basis. The retrieved  
23 aerosol optical properties have been used in numerous air quality studies as well as studies  
24 related to the effect of airborne particles on various climatic parameters (e.g. radiation, clouds,  
25 precipitation, etc.). The  $1^\circ \times 1^\circ$  daily gridded level-3 dataset is primarily used for global as well  
26 as regional studies while the single pixel level-2 data with a 10 km resolution (at nadir) are  
27 mostly used for regional and local scale studies. Nevertheless, the use of the coarse resolution  
28 MODIS data has predominated even in regional studies. The reasons for this could be the  
29 smaller file size which makes their processing and storage easier or the fact that they are easily  
30 accessible through user-friendly data bases which also allow for a very basic analysis like e.g.  
31 NASA's GIOVANNI website (<http://giovanni.gsfc.nasa.gov/giovanni/>) (Acker and Leptoukh,  
32 2007).

1 The same holds for studies focusing on the Mediterranean Basin, an area which is considered of  
2 particular sensitivity as far as air pollution and climate change is concerned (Lelieveld et al.,  
3 2002; Giorgi, 2006). The Mediterranean basin is one of the regions with the highest aerosol  
4 optical depths (AODs) in the world (Husar et al., 1997; Ichocku et al., 2005; Papadimas et al.,  
5 2008), causing significant climate forcing especially in summer, which is characterized by low  
6 cloudiness and high incoming solar radiation levels (Papadimas et al., 2012; Alexandri et al.,  
7 2015). The Mediterranean is also recognized as a crossroads between three continents where  
8 aerosols of various types accumulate (Lelieveld et al., 2002). Marine aerosols from the  
9 Mediterranean Sea and even the Atlantic Ocean combine with aerosols from continental Europe  
10 (urban and rural), dust particles transported from the Sahara Desert and Middle East as well as  
11 biomass burning aerosols from occasional wild fires and agricultural burning (Lelieveld et al.,  
12 2002). Specifically, as discussed in Hatzianastassiou et al. (2009), Eastern Mediterranean, the  
13 region under investigation here, is located at a "key" point of this crossroads. There is a  
14 significant number of ground and satellite-based studies on the abundance and optical properties  
15 of tropospheric aerosols in the area; however, these studies are either focused on specific spots  
16 or used a coarse spatial and temporal resolution.

17 The ground-based instrumentation used in studies focusing on the aerosol load and optical  
18 properties over the Eastern Mediterranean includes active and passive sensors such as Lidars  
19 (e.g. Papayannis and Balis, 1998; Balis et al., 2004; Papayannis et al., 2005, 2009; Amiridis et  
20 al., 2005, 2009; Mamouri et al., 2013; Kokkalis et al., 2013; Nisantzi et al., 2015), Cimel  
21 sunphotometers (e.g. Israelevich et al., 2003; Kubilay et al., 2003; Derimian et al., 2006;  
22 Kalivitis et al., 2007; Kelektsoğlu and Rapsomanikis, 2011; Nikitidou and Kazantzidis, 2013),  
23 Brewer spectrophotometers (e.g. Kazadzis et al., 2007; Koukouli et al., 2010), Multi-Filter  
24 Radiometers (e.g. Gerasopoulos et al., 2009, 2011; Kazadzis et al., 2014), ceilometers (e.g.  
25 Tsaknakis et al., 2011), Microtops sunphotometers (e.g. El-Metwally and Alfaro, 2013), etc.  
26 However, these and other studies not referenced here either refer to specific spots with the  
27 majority of the ground stations being situated in large urban centers (e.g. Athens, Thessaloniki,  
28 Cairo) or to specific events (e.g. Sahara dust intrusions, biomass burning events, etc.).

29 On the other hand, AOD and other aerosol optical properties have been studied over the greater  
30 Eastern Mediterranean region based on data from Meteosat (Moulin et al., 1998), SeaWiFS  
31 (Koren et al., 2003; Antoine and Nobileau, 2006; Mélin et al., 2007; Nabat et al., 2013), TOMS  
32 (Alpert and Ganor, 2001; Israelevich et al., 2002; Koukouli et al. 2006; Hatzianastassiou et al.,  
33 2009; Koukouli et al., 2010, Israelevich et al., 2012, Kaskaoutis et al., 2012a; Nabat et al., 2013;

1 Gkikas et al., 2013, 2014; Varga et al., 2014), MODIS Terra and Aqua (Barnaba and Gobbi,  
2 2004; Papayannis et al., 2005; Kaskaoutis et al., 2007, 2008, 2010, 2011, 2012a,b,c,d;  
3 Kosmopoulos et al., 2008; Papadimas et al., 2008, 2009; Rudich et al., 2008; Carmona and  
4 Alpert, 2009; Karnieli et al., 2009; Gkikas et al., 2009, 2013; Hatzianastassiou et al., 2009; El-  
5 Metwally et al., 2010; Koukouli et al., 2010; Kanakidou et al., 2011; Gerasopoulos et al., 2011;  
6 de Meij and Lelieveld, 2011; Marey et al., 2011; de Meij et al., 2012; Nabat et al., 2012, 2013;  
7 Nikitidou and Kazantzidis, 2013; Athanasiou et al., 2013; Benas et al., 2011, 2013; Sorek-  
8 Hamer et al., 2013; Kabatas et al., 2014; Kourtidis et al., 2014; Mishra et al., 2014; Flaounas et  
9 al., 2015; Kloog et al., 2015), OMI/AURA (Kaskaoutis et al., 2010; El-Metwally et al., 2010;  
10 Marey et al., 2011; Kaskaoutis et al., 2012b,c, Gkikas et al., 2013, 2014; Sorek-Hamer et al.,  
11 2013; Varga et al., 2014; Flaounas et al., 2015), CALIOP/CALIPSO (Amiridis et al., 2009,  
12 2013, Mamouri et al., 2009; Marey et al., 2011; Kaskaoutis et al., 2012c; de Meij et al., 2012;  
13 Nabat et al., 2012, 2013; Mamouri and Ansmann, 2015), MISR/Terra (Kanakidou et al., 2011;  
14 Marey et al., 2011; de Meij and Lelieveld, 2011; de Meij et al., 2012; Nabat et al., 2013;  
15 Kabatas et al., 2014; Abdelkader et al., 2015) as well as NOAA/AVHRR, MERIS/ENVISAT,  
16 AATSR/ENVISAT, PARASOL/POLDER, MSG/SEVIRI, and Landsat satellite data (see  
17 Retalis and Sifakis, 2010; Nabat et al., 2013; Benas et al., 2013; Sifakis et al., 2014). To our  
18 knowledge, these studies comprise the majority of works focusing on tropospheric aerosols over  
19 the Eastern Mediterranean by means of satellite remote sensing, published in peer reviewed  
20 journals the last ~ 15 years. As shown in Fig. 4 of this work, the publication rate of satellite-  
21 based studies focusing on the Eastern Mediterranean aerosols nearly doubled every three years  
22 during the period 1997-2014 which is indicative of the increasing scientific interest in the area.  
23 In a very large fraction of the satellite-based studies referenced above, the used data are either of  
24 coarse mode (usually  $1^\circ$  which is ~ 100 km for the mid-latitudes) or focus on specific spots for  
25 validation purposes. In a few cases, high resolution data were used in spatiotemporal studies;  
26 however, either these studies are restricted over surfaces covered by water or examine a short  
27 period only. For example, Moulin et al. (1998) investigated the dust AOD patterns over the  
28 oceanic areas of the Mediterranean Basin at a resolution of  $35 \times 35 \text{ km}^2$  for a period of 11 years  
29 (1984-1994) using Meteosat observations. A 7-year climatology (1998-2004) of total and dust  
30 AOD for the same regions at a resolution of  $0.16^\circ \times 0.16^\circ$  was compiled by Antoine and  
31 Nobileau (2006) using observations from SeaWIFS. Mélin et al. (2007) merged SeaWIFS and  
32 MODIS data and presented high resolution AOD patterns ( $2 \times 2 \text{ km}^2$ ) for May 2003. As far as  
33 MODIS is concerned, only Barnaba and Gobbi (2004) presented a high resolution ( $0.1^\circ \times 0.1^\circ$ )

1 spatiotemporal analysis for a period of 1 year (2001) over sea only. In a recent paper,  
2 Athanasiou et al. (2013) presented in detail a method for compiling a 0.5-degree resolution  
3 AOD gridded dataset using level-2 MODIS Terra data for the greater region of Greece (2000-  
4 2008). However, the spatial resolution they used ( $\sim 50$  km) is not high enough to reveal local  
5 sources (e.g. cities, islands, river banks, etc.). Overall, there has not been so far any detailed  
6 high resolution spatiotemporal study of the AOD over the Eastern Mediterranean.

7 In this paper, the  $AOD_{550}$  spatiotemporal variability over the Eastern Mediterranean ( $30^{\circ}\text{N}$ -  
8  $45^{\circ}\text{N}$ ,  $17.5^{\circ}\text{E}$ - $37.5^{\circ}\text{E}$ ) is presented at a high spatial resolution ( $0.1^{\circ} \times 0.1^{\circ}$ ) based on MODIS  
9 Terra and Aqua observations. Level-2 MODIS data are used for the compilation of a 0.1-degree  
10 gridded dataset which is validated against ground-based observations. In order to calculate the  
11 contribution of different aerosol types to the total AOD, the MODIS data were analyzed  
12 together with other satellite data, ERA-Interim and MACC reanalysis data and the Goddard  
13 Chemistry Aerosol Radiation and Transport (GOCART) model using an algorithm optimized  
14 for the surface properties of the Eastern Mediterranean region. The different datasets used in  
15 this research are presented in detail in Sect. 2 while a detailed description of the method is given  
16 in Sect. 3. Sect. 4 includes the results from the MODIS validation procedure, the annual and  
17 seasonal variability of  $AOD_{550}$  over the region with a discussion on the local aerosol sources  
18 and the differences between Terra and Aqua, and the annual and seasonal contribution of  
19 different aerosol types to the total  $AOD_{550}$ . Finally, in Sect. 5, the main conclusions of the paper  
20 are presented along with a short discussion on how these results could contribute to future  
21 studies in the area.

22

## 23 **2 Observations, reanalysis data and model simulations**

### 24 **2.1 MODIS Terra and Aqua satellite observations**

25 The main data used in this work come from the level-2 MODIS Terra (MOD04\_L2) and  
26 MODIS Aqua (MYD04\_L2) Collection 051 dataset and have been acquired through NASA's  
27 Level 1 and Atmosphere Archive and Distribution System (LAADS)  
28 (<http://ladsweb.nascom.nasa.gov>). MODIS Terra and Aqua have a daytime equator crossing  
29 time at 10:30 LT (morning) and 13:30 LT (noon), respectively. MODIS instruments with a  
30 viewing swath of 2330 km measure backscattered radiation at 36 spectral bands between  
31 0.415 and 14.235  $\mu\text{m}$  with a spatial resolution of 250, 500 and 1000 m, providing a nearly  
32 global coverage on a daily basis. Aerosol optical properties for the standard MODIS aerosol  
33 product are retrieved using two different "Dark Target" (DT) algorithms. The one is used over

1 land surfaces (Kaufman et al., 1997; Levy et al., 2007a, b; Remer et al., 2005; Levy et al.,  
2 2010) and the other over oceanic regions (Tanré et al., 1997; Levy et al., 2003; Remer et al.,  
3 2005). The "Deep Blue" algorithm (DB) (Hsu et al., 2004; Hsu et al., 2006) has been used for  
4 retrievals over bright land surfaces (e.g. deserts) where the DT algorithm fails. Only recently,  
5 updates to the algorithm allowed for extending the spatial coverage of the DB aerosol product  
6 over all land areas (Hsu et al., 2013; Sayer et al., 2013; 2014). AERONET Cimel  
7 sunphotometric measurements have been extensively used for the validation of the MODIS  
8 over-land and over-ocean products (e.g. Chu et al., 2002; Remer et al., 2002; Remer et al.,  
9 2005; Levy et al., 2010; Shi et al., 2013).

10 In this work, AOD<sub>550</sub> over both land and sea and the Fine Mode Ratio (FMR<sub>550</sub>) over sea from  
11 Collection 051 were used at a spatial resolution of 10 km (at nadir). The uncertainty of the  
12 MODIS aerosol optical depth has been estimated at  $\pm(0.05+0.15AOD)$  over land (Chu et al.,  
13 2002; Levy et al., 2010) and  $\pm(0.03+0.05AOD)$  over ocean (Remer et al., 2002) relative to the  
14 AERONET AOD. Specifically, for the DT data used in this work only high quality retrievals  
15 are used over land. This means that the data have a Quality Assurance Confidence (QAC) flag  
16 equal to 3 (high confidence). For retrievals over sea we use data with a QAC flag of 1  
17 (marginally good), 2 (good) and 3 (see Levy et al., 2009 for details). The pre-launch  
18 uncertainty of FMR<sub>550</sub> is  $\pm 30\%$  over ocean (Remer et al., 2005) while over land this  
19 parameter is by no means trustworthy and should only be used in qualitative studies (e.g. see  
20 Georgoulas and Kourtidis, 2011). In cases where DT algorithm does not provide products  
21 over land, especially over bright arid and semi-arid regions of North Africa, AOD<sub>550</sub> values  
22 from the DB algorithm are used in our work. The expected uncertainty of the DB product  
23 used here is  $\pm(0.05+0.2AOD)$  relative to the AERONET AOD (Hsu et al., 2006). The  
24 analyzed datasets cover the period from 3/2000 to 12/2012 for Terra and from 7/2002 to  
25 12/2012 for Aqua MODIS covering the region of the Eastern Mediterranean. The Collection  
26 051 DB data for Terra are available only until 12/2007 due to calibration issues; nevertheless,  
27 these data are carefully used within our analysis to get a complete image of the aerosol load  
28 over the region.

29

## 30 **2.2 AERONET ground-based observations**

31 For the evaluation of the MODIS AOD<sub>550</sub>, version 2.0 level 2.0 high quality cloud screened  
32 data from 13 AERONET Cimel network ground stations in the region of the Eastern  
33 Mediterranean have been acquired (<http://aeronet.gsfc.nasa.gov>). The stations were selected

1 such that their operation period covers at least 2 years and there are at least 100 common days  
2 of co-localized AERONET and MODIS observations. AERONET Cimel sunphotometers  
3 measure solar radiation every 15 minutes within the spectral range from 340 to 1020 nm  
4 (Holben et al., 2001). The spectral measurements allow for the retrieval of columnar aerosol  
5 properties (see Holben et al., 1998; Dubovik and King, 2000; Dubovik et al., 2000, 2002).  
6 The AERONET AOD uncertainty is in the order of 0.01-0.02 (Eck et al., 1999), being larger  
7 at shorter wavelengths. Here, we use quadratic fits on a log-log scale to interpolate the  
8 AERONET data (AODs at 440, 500, 675 and 870 nm) to the MODIS band-effective  
9 wavelength of 550 nm (Eck et al., 1999; Levy et al., 2010). So, we can directly compare the  
10 MODIS AOD<sub>550</sub> retrievals against AERONET observations. Simultaneous measurements of  
11 the Ångström Exponent (AE) for the spectral range 440-870 nm (AE<sub>440-870</sub>) from the 13  
12 AERONET stations mentioned above were also utilized in this work in order to account for  
13 days with dust dominance. The uncertainty of the AE is significantly higher than the AOD  
14 uncertainty, especially under low-AOD conditions. Li et al. (2014) found that the uncertainty  
15 for a typical Northern Hemispheric AERONET station (GSFC) is  $\sim 0.6$  during winter when  
16 AODs are significantly lower compared to summer ( $\sim 0.15$ ).

17

### 18 **2.3 LIVAS CALIOP/CALIPSO dust climatology**

19 Dust aerosol optical depths at 532 nm (AOD<sub>532</sub>) from CALIOP/CALIPSO (Cloud-Aerosol  
20 Lidar with Orthogonal Polarization instrument aboard Cloud-Aerosol Lidar and Infrared  
21 Pathfinder Satellite Observations satellite) at a resolution of  $1^\circ \times 1^\circ$  are also used here for the  
22 period 2007-2012. CALIPSO measures cloud and aerosol properties flying at a 705 km sun  
23 synchronous polar orbit with a 16 day repeat cycle and an equator-crossing time close to that  
24 of the Aqua satellite (13:30 LT). The dust product used here comes from a Saharan-dust-  
25 optimized retrieval scheme that was developed within the framework of the LIVAS (Lidar  
26 Climatology of Vertical Aerosol Structure for Space-Based LIDAR Simulation Studies)  
27 project (Amiridis et al. 2015) and has been presented in detail in Amiridis et al. (2013). In  
28 brief, the LIVAS dust product is optimized for Europe by applying a lidar ratio of 58 sr  
29 instead of 40 sr to Level 2 dust related backscatter products. This correction results to an  
30 improvement of the AOD<sub>532</sub> product. Comparison against spatially and temporally co-located  
31 AERONET observations (Amiridis et al., 2013) returned an absolute bias of  $\sim -0.03$ . The  
32 corresponding reported biases for the original CALIPSO data are significantly higher ( $\sim -$   
33 0.10). The bias is even lower when compared against MODIS satellite-based observations.

1 Other improvements of this product are related to the use of a new methodology for the  
2 calculation of pure dust extinction from dust mixtures and the application of an averaging  
3 scheme that includes zero extinction values for the non-dust aerosol types detected. Overall,  
4 this product (hereafter denoted as LIVAS dust product) exhibits better agreement with  
5 observations from MODIS and AERONET and simulations from the BSC-DREAM8b dust  
6 model over North Africa and Europe than the standard CALIPSO data hence being an ideal  
7 tool for the evaluation of other satellite-based products.

8

#### 9 **2.4 Earth Probe TOMS and OMI satellite observations**

10 For this work, UV Aerosol Index (AI) data (Herman et al., 1997) from the Earth Probe TOMS  
11 (Total Ozone Mapping) spectrometer aboard Earth Probe for the period 1/2000-9/2004 at a  
12 resolution of  $1^\circ$  (latitude)  $\times$   $1.25^\circ$  (longitude) and the OMI (Ozone Monitoring Instrument)  
13 sensor aboard EOS AURA for the period 10/2004-12/2012 at a resolution of  $1^\circ \times 1^\circ$  were  
14 acquired through the GIOVANNI web database (<http://giovanni.gsfc.nasa.gov/giovanni/>).  
15 Earth Probe TOMS continued the record of the first three TOMS instruments aboard Nimbus-  
16 7, Meteor-3 and ADEOS flying in a sun synchronous orbit at an altitude of 740 km with an  
17 instantaneous field of view size of  $39 \times 39 \text{ km}^2$  at nadir. The instrument had an ascending  
18 node equator crossing time at 12:00 LT covering 85 % of the globe on a daily basis from  
19 7/1996 until 12/2005. The satellite was originally set to a 500 km sun synchronous orbit but  
20 was set to its final orbit after the failure of ADEOS satellite in 6/1997. OMI is a UV/VIS  
21 nadir solar backscatter spectrometer (Levelt et al., 2006) that continues the long TOMS  
22 record. OMI flies in a sun synchronous polar orbit at an altitude of 705 km with an ascending  
23 node equator crossing time at 13:45 LT. Its 2600 km viewing swath allows for almost daily  
24 global coverage while the spatial resolution of the instrument is  $13 \times 24 \text{ km}^2$  at nadir. The AI  
25 (also known as Absorbing Aerosol index) which is calculated by the two instruments  
26 constitutes a qualitative indicator of the presence of UV absorbing aerosols in the atmosphere  
27 such as biomass burning and dust (Torres et al., 1998). Positive AI values generally represent  
28 absorbing aerosols while small or negative values represent non-absorbing aerosols. The  
29 Version 8 algorithm is applied to spectral measurements from both TOMS and OMI sensor to  
30 produce a consistent long-term AI timeseries (Li et al., 2009). AI is calculated from the  
31 difference in surface reflectivity derived from the 331.2 and 360 nm measurements exhibiting  
32 an uncertainty of  $\pm 30 \%$  (Torres et al., 2007).

33



## 2.5 ERA-Interim reanalysis data

Wind speed (ws) data at 10 m above surface from the ERA-Interim reanalysis (Dee et al., 2011) are used for 9:00 and 12:00 UTC on a daily basis for the period 2000-2012. We use 9:00 and 12:00 UTC data in order to be closer to the Terra and Aqua overpass time in the area, respectively. The various ERA-Interim reanalysis fields are produced by ECMWF's Integrated Forecast System (IFS) assimilating satellite and ground-based observations. The system includes a 4-D variational analysis with a 12-hour analysis window. The spatial resolution of the ERA-Interim data is  $\sim 79$  km with 60 vertical levels from the surface up to 0.1 hPa while the data can be acquired at various resolutions (in this work  $1^\circ \times 1^\circ$ ) through ECMWF's website (<http://apps.ecmwf.int/datasets/data/interim-full-daily/>). Over ocean, the 10 m ERA-interim wind speed exhibits a bias of less than -0.5 m/s compared to quality-controlled in situ observations on a global scale (Dee et al., 2011). Specifically, for the region of the Eastern Mediterranean examined here, the 10 m ERA-interim wind speed exhibits a bias of -0.96 m/s (-16 %) compared to satellite-based observations from QuikSCAT (Hermann et al., 2011).

## 2.6 MACC reanalysis data

The daily MACC total and dust AOD<sub>550</sub> data for the period 2003-2012 come from the aerosol analysis and forecast system of ECMWF which consists of a forward model (Morcrette et al., 2009) and a data-assimilation module (Benedetti et al., 2009). AOD<sub>550</sub> measurements from the two MODIS instruments aboard Terra and Aqua are assimilated by the MACC forecasting system through a 4D-Var assimilation algorithm to produce the aerosol analysis, leading to an improved AOD representation compared to observations (see Benedetti et al., 2009; Mangold et al., 2011). Five aerosol species are included within MACC, namely, mineral dust, sea salt, sulfates, black carbon and organic matter. Three different size bins are used for mineral dust and sea salt particles while the black carbon and organic material are distributed to a hydrophilic and a hydrophobic mode. Dust and sea salt emissions are given as a function of surface wind speed, while the emissions of the other species are taken from inventories. The spatial resolution of the MACC reanalysis data is  $\sim 79$  km with 60 vertical levels from the surface up to 0.1 hPa and can be acquired through: <http://apps.ecmwf.int/datasets/data/macc-reanalysis/> for the period 2003-2012. The MACC total and dust AODs have been evaluated against ground and satellite-based observations (see Elguindi et al., 2010; Bellouin et al., 2013; Inness et al., 2013; Cesnulyte et al., 2014; Cuevas et al., 2015) showing that the MACC

1 aerosol products generally capture well the daily, seasonal and interannual variability of  
2 aerosols. As discussed in Bellouin et al. (2013) the uncertainties of MACC total AOD<sub>550</sub> (~  
3 0.03) and dust AOD<sub>550</sub> (~ 0.014) arise from uncertainties in the MODIS retrievals which are  
4 assimilated into the model and errors in the forward modeling of total and component AODs.  
5

## 6 **2.7 GOCART data**

7 Daily total and dust AOD<sub>550</sub> data from the GOCART chemistry-aerosol-transport model  
8 simulations (version 006) are used in this study for the period 2000-2007. The GOCART  
9 model (see Chin et al., 2000, 2002, 2004, 2007; Ginoux et al., 2001, 2004) uses the  
10 assimilated meteorological fields of the Goddard Earth Observing System Data Assimilation  
11 System (GEOS DAS) which are generated by the Goddard Global Modeling and Assimilation  
12 Office (GMAO). The data which are used were acquired from an older version of NASA's  
13 GIOVANNI web database (<http://disc.sci.gsfc.nasa.gov/giovanni/>) and come from a  
14 simulation implemented at a spatial resolution of 2° (latitude) x 2.5° (longitude) with 30  
15 vertical sigma layers (Chin et al., 2009). The model includes physicochemical processes of  
16 major tropospheric aerosol components (sulfates, dust, black carbon, organic carbon, sea salt)  
17 and precursor gases (SO<sub>2</sub> and dimethylsulfide) incorporating various atmospheric processes.  
18 The total AOD<sub>550</sub> from GOCART compared to ground-based observations from the  
19 AERONET exhibits a relative mean bias [mean(GOCART)/mean(AERONET)] of 1.120,  
20 1.135 and 0.959 over Europe, North Africa and for the whole globe, respectively.  
21

## 22 **2.8 Ancillary data**

23 Apart from the main datasets presented above, three additional datasets were used in order to  
24 support our findings. OMI/AURA daily gridded (Bucsela et al., 2013) tropospheric NO<sub>2</sub>  
25 columnar data (OMNO2d version 2.1) at a spatial resolution of 0.25° x 0.25° were acquired  
26 from NASA's GIOVANNI web database (<http://giovanni.gsfc.nasa.gov/giovanni/>) for the  
27 period 2005-2012. The quality checked data used in this work correspond to sky conditions  
28 where cloud fraction is less than 30 %. Planetary boundary layer (PBL) SO<sub>2</sub> daily gridded  
29 columnar data (OMSO2e version 1.1.7) were also acquired from GIOVANNI for the same  
30 period. The OMSO2e gridded data (0.25° x 0.25°) used in this work are produced from best  
31 level-2 pixel data, screened for OMI row anomaly and other data quality flags. The PBL SO<sub>2</sub>  
32 column retrievals are produced with an algorithm based on principal component analysis  
33 (PCA) of the OMI radiance data (Li et al., 2013). Finally, monthly precipitation data from the

1 3B43 TRMM and Other Sources Monthly Rainfall Product (version 7) at a spatial resolution  
2 of  $0.25^\circ \times 0.25^\circ$  for the period 2000-2012 were obtained from GIOVANNI. This dataset is  
3 derived from 3-hourly precipitation retrievals from the Precipitation Radar (PR), the TRMM  
4 Microwave Imager (TMI) and the Visible and Infrared Scanner (VIRS) aboard the TRMM  
5 (Tropical Rainfall Monitoring Mission) satellite merged with other satellite-based  
6 precipitation data and the Global Precipitation Climatology Centre (GPCC) rain gauge  
7 analysis (Huffman et al., 2007).

### 8 9 **3 Methodology**

#### 10 **3.1 Compiling a MODIS $0.1^\circ \times 0.1^\circ$ gridded dataset**

11 To investigate the spatial and temporal variability of aerosols over the Eastern Mediterranean  
12 we first created a  $0.1^\circ \times 0.1^\circ$  daily gridded aerosol dataset using single pixel level-2 AOD<sub>550</sub> and  
13 FMR<sub>550</sub> data from MODIS Collection 051. The same resolution has been utilized in previous  
14 studies (e.g. Barnaba and Gobbi, 2004) in the region; however, without reporting on the  
15 gridding methodology followed. In this work we present a gridding methodology that could be  
16 used as a reference for future regional studies. The methodology has been successfully applied  
17 in the past on level-2 MODIS Terra data in different cases studies, e.g. in order to examine the  
18 weekly cycle patterns of AOD<sub>550</sub> over the region of Central Europe and the aerosol load  
19 changes observed over a cement plant in Greece due to changes in the deposition practices of  
20 the primary materials (see Georgoulas and Kourtidis, 2012; Georgoulas et al., 2012; Kourtidis  
21 et al., 2014). In the following lines we proceed to a detailed description of the method  
22 underlining the potential of being used in detailed quantitative studies like this one.

23 First, a  $0.1^\circ \times 0.1^\circ$  resolution grid covering the Eastern Mediterranean ( $30^\circ\text{N}$ - $45^\circ\text{N}$ ,  $17.5^\circ\text{E}$ -  
24  $37.5^\circ\text{E}$ ) is defined which corresponds to 30000 grid cells. As already mentioned in Sect. 2.1,  
25 only level-2 single pixel AOD<sub>550</sub> measurements with a QAC flag of 3 and a QAC flag greater  
26 than 0 were used over land and over sea, respectively, to ensure the high quality of the data.  
27 Pixels are attributed to a specific grid cell if their center falls within a  $25 \times 25 \text{ km}^2$  square  
28 window around the grid cell (see Fig. S1 in the Supplement). These pixels are then used for the  
29 calculation of daily averages. As shown in Figure S1, a grid cell of  $0.1^\circ$  ( $\sim 10 \text{ km}$ ) is as big as  
30 the centre of a large Mediterranean city like Thessaloniki, Northern Greece ( $\sim 1$  million  
31 inhabitants). The procedure was followed separately for MODIS Terra and Aqua data. In cases  
32 of grid cells with no DT MODIS observations, data from the DB algorithm were used (over

1 bright arid and semi-arid regions of North Africa) constituting only a small part of the gridded  
2 dataset.

3 The size of the gridding window was selected following Koukouli et al. (2007). They used both  
4 10 and 25-km windows showing that the latter allows for the inclusion of more data points  
5 without undermining the ability of monitoring accurately the aerosol load over a specific spot.  
6 In addition, in cases of urban sites, a window of 25 km allows for the inclusion of pixels from  
7 the surrounding non-urban surfaces where the MODIS surface reflectance parameterization is  
8 better (Levy et al., 2010). The size of each MODIS pixel is 10 km at nadir, but at the swath  
9 edges, it may become 2-3 times larger. Hence, ideally the maximum number of pixels that could  
10 be used in the daily averaging is nine. The overlap between the windows of neighbouring grid  
11 cells does not affect the representativeness of the dataset over each grid cell. Aerosols are  
12 transported by air masses throughout the day and thus the aerosol load in neighbouring grid  
13 cells is not expected to be completely independent.

14 In order to make sure that the use of a 25-km gridding window is optimal for capturing local  
15 pollution sources we repeated the same procedure for bigger gridding windows (50-km, 75km  
16 and 100-km) using MODIS Terra AOD<sub>550</sub> data for the year 2004. Numerous aerosol hot spots  
17 cannot be seen as the gridding window becomes bigger and there is a significant smoothing of  
18 the aerosol patterns mainly over land (Fig. S2). The use of the MODIS gridded dataset in the  
19 detection of local aerosol hot spots is discussed in more detail in Sect. 4. In addition, we  
20 conducted a detailed validation of the MODIS data against sunphotometric data from a total of  
21 13 AERONET stations in the region (see Fig. 1). The validation procedure was repeated several  
22 times for different spatial collocation windows which were equal to the windows used for the  
23 gridding procedure (i.e. 25, 50, 75 and 100-km) and for different data quality criteria. The  
24 results of the validation procedure are presented in Sect. 4.1 while part of them is given in the  
25 Supplement of this manuscript (see Table S2). Overall, it is shown that the gridding  
26 methodology followed here offers the best compromise for studying the spatial variability of  
27 aerosols on a regional or local scale, preserving at the same time the representativeness of the  
28 real aerosol load over each specific spot.

29 In order to generalize our results, nine different sub-regions (Fig. 1) were selected apart from  
30 the three basic regions of interest, namely, the whole Eastern Mediterranean (EMT) and the land  
31 (EML) and oceanic (EMO) areas of the region. The selection was done mainly taking into  
32 account geographical but also land type and land use criteria. The four sub-regions that  
33 correspond to the land regions of the Eastern Mediterranean are the Northern Balkans Land

1 (NBL), the Southern Balkans Land (SBL), the Anatolia Land (ANL) and the Northern Africa  
2 Land (NAL) region while the five sub-regions that correspond to the oceanic regions are the  
3 Black Sea Oceanic (BSO), the North-Western Oceanic (NWO), the South-Western Oceanic  
4 (SWO), the North-Eastern Oceanic (NEO) and the South-Eastern Oceanic (SEO) region. Mean  
5 values of the total  $AOD_{550}$  from the Terra and Aqua MODIS are reported for each one of the  
6 three basic regions of interest and their nine sub-regions in Sect. 4.

## 8 **3.2 Contribution of different aerosol types to $AOD_{550}$**

### 9 **3.2.1 Sea**

10 In order to quantify the contribution of different types of aerosols to the total  $AOD_{550}$  we  
11 followed a different approach for sea and land. This is due to the lack of reliable  $FMR_{550}$   
12 retrievals over land (e.g. see Levy et al., 2010; Georgoulias and Kourtidis, 2011) which are  
13 crucial for the algorithms used in this work. Over the sea we utilize wind speed data at 10 m  
14 above surface from the ERA-Interim reanalysis, AI data from TOMS and OMI along with  
15  $AOD_{550}$  and  $FMR_{550}$  from the MODIS Terra and Aqua gridded datasets presented above. All the  
16 datasets were brought to the same 0.1 degree spatial resolution as MODIS by using bilinear  
17 interpolation. In the case of TOMS and OMI we used monthly mean AI data following Bellouin  
18 et al. (2008) in order to avoid gaps especially during the TOMS period.

19 In general, the algorithm used over the oceanic regions (see Fig. 2) is similar with the one  
20 presented in Bellouin et al. (2008). First, the marine  $AOD_{550}$  ( $\tau_m$ ) is calculated from near surface  
21 wind speed using a linear relation which has been obtained from ground-based studies over  
22 pollution free oceanic regions. Bellouin et al. (2008) use the linear relation of Smirnov (2003).  
23 Then, if  $\tau_m$  is greater or equal than  $AOD_{550}$  it is assumed that there are marine particles only  
24 over this region. If  $\tau_m$  is smaller than  $AOD_{550}$  a decision tree is followed which is first based on  
25  $FMR_{550}$  and then on AI in order to reach a conclusion about the type of aerosols that account for  
26  $AOD_{550}$ . If  $FMR_{550}$  is smaller than the critical value of 0.35 and AI is greater than or equal to a  
27 critical value it is assumed that there are both marine aerosols ( $\tau_m$ ) and dust ( $\tau_d=AOD_{550}-\tau_m$ ),  
28 while, if AI is smaller than this critical value it is assumed that there are marine aerosols only.  
29 The AI critical value is equal to 1 in Bellouin et al. (2008). If  $FMR_{550}$  is greater than or equal to  
30 0.83 it is assumed that there are both anthropogenic ( $\tau_a=AOD_{550}-\tau_m$ ) and marine aerosols ( $\tau_m$ ).  
31 In the occasion of having a  $FMR_{550}$  equal to 0.35 or greater than 0.35 but smaller than 0.83 one  
32 has to take again AI into consideration. If AI is less than the critical value it is assumed that  
33 there are marine aerosols ( $\tau_m$ ) only while in the opposite occasion it is assumed that all the three

1 types of aerosols that can be defined over oceanic regions by this algorithm, namely, dust  
2 [ $\tau_d=(1-FMR_{550})(AOD_{550}-\tau_m)$ ], anthropogenic [ $\tau_a=FMR_{550}(AOD_{550}-\tau_m)$ ] and marine aerosols ( $\tau_m$ )  
3 are present. One should keep in mind that all the biomass burning aerosols are classified as  
4 anthropogenic by this method.

5 In this work, we proceeded to a "fine-tuning" of the algorithm for the region of the Eastern  
6 Mediterranean. First, we applied the algorithm on MODIS Terra data using the same equations  
7 and critical values as in Bellouin et al. (2008). The results showed that the original Bellouin et  
8 al. (2008) method might be valid for global studies but for a "closed" sea like the Mediterranean  
9 the method leads to a large overestimation of sea salt AODs and therefore underestimation of  
10 dust and anthropogenic aerosol AODs. Indicative of this situation is Fig. S3 in the Supplement  
11 where we present the relative contribution of dust, marine and anthropogenic aerosols per  
12 month over the oceanic regions of the Eastern Mediterranean as calculated using the original  
13 Bellouin et al. (2008) method. It is shown that the marine contribution is several times higher  
14 than the values reported for the Mediterranean Basin in previous studies (e.g. see Nabat et al.,  
15 2012). Evaluation of the algorithm was done using dust AOD<sub>532</sub> data from the LIVAS  
16 CALIOP/CALIPSO product. From LIVAS we only use the high quality Sahara dust product as  
17 a reference and not other aerosol type retrievals (e.g. marine aerosols) since the dust retrievals  
18 from CALIOP/CALIPSO are by far the most reliable (e.g. Burton et al., 2013). We performed  
19 several tests by changing the linear relation that connects  $\tau_m$  with near surface wind speed and  
20 the AI critical values and compared each time the dust AOD<sub>550</sub> seasonal variability with the  
21 LIVAS AOD<sub>532</sub> seasonal variability for the sea covered sub-regions of the Eastern  
22 Mediterranean. Results from this algorithm-tuning procedure can be found in Figs. S4e-i of the  
23 Supplement where one can also see the underestimation of dust AOD<sub>550</sub> from the original  
24 Bellouin et al. (2008) algorithm.

25 The linear relation given in Kaufman et al. (2005) was finally selected ( $\tau_m=0.007ws+0.02$ ). The  
26 2000-2012 average wind speed over the sea for the region of the Eastern Mediterranean is  $\sim 5.3$   
27 m/s. Kaufman et al. (2005) reduced the offset in the linear relation of Smirnov (2003) from 0.06  
28 to 0.02 to fit the average baseline AOD of 0.06 for the typical wind speed of 6 m/s. In addition,  
29 our tests showed that an AI critical value of 1 performs well over the region of the Eastern  
30 Mediterranean. The results did not change significantly when using other AI thresholds (e.g. 0.5  
31 which is suggested in Jones and Christopher, 2011) and therefore we decided to adopt 1 as the  
32 AI critical value. Another test, following the example of other studies (see Lehahn et al., 2010),  
33 was to assume that for wind speed less than 5 m/s there is very little or no sea-spray particle

1 production (limited bursting of entrained air bubbles associated with whitecap formation). In  
2 this case,  $\tau_m$  is stable, equal to the offset of the linear relation between  $\tau_m$  and wind speed which  
3 is indicative of the background sea salt AOD<sub>550</sub>. However, this test reveals that the effect of  
4 assuming stable  $\tau_m$  for wind speed less than 5 m/s is insignificant and therefore we selected to  
5 follow the Kaufman et al. (2005) linear relationship for the whole wind speed range. As shown  
6 in Figs. S4e-i, the seasonal variability when applying our modified algorithm over oceanic  
7 regions is very close to the LIVAS dust AOD<sub>532</sub> especially for the months with lower dust load  
8 (June-January). It is also shown that dust AODs from this algorithm are closer to the LIVAS  
9 dust product than dust AODs from MACC reanalysis do. The slight overestimation of dust  
10 AOD or the shift of the maximum dust load we observe for the period of high dust loads in the  
11 region (February-May) is probably connected to the narrow swath and the 16-day time of  
12 CALIPSO which means that several dust events might be not observed by the CALIOP  
13 instrument contrary to MODIS which has a daily coverage.

14

### 15 **3.2.2 Land**

16 As already mentioned in the previous paragraph a different approach is followed over the land  
17 regions of the Eastern Mediterranean due the low confidence on the MODIS FMR<sub>550</sub> and  
18 Ångström exponent retrievals over land compared to that over ocean (e.g. see Levy et al., 2010;  
19 Georgoulas and Kourtidis, 2011). This limitation does not allow us to distinguish the  
20 contribution of fine and coarse mode aerosols in terms of AOD<sub>550</sub>. In this case, we choose to use  
21 daily model fields of the dust contribution to the total AOD (here MACC reanalysis and  
22 GOCART). We follow a method similar with the one presented in Bellouin et al. (2013).  
23 Specifically, we calculate the dust AOD<sub>550</sub> by scaling the MODIS AOD<sub>550</sub> data with the MACC  
24 or GOCART dust/total AOD<sub>550</sub> ratios [ $f_d = \tau_{d(model)}/\tau_{(model)}$ ] on a daily basis.

25 Since the MACC data are available only from 2003 to 2012, in order to take advantage of the  
26 full MODIS dataset (3/2000-12/2012), data from the GOCART model were used for the period  
27 2000-2002. The GOCART data were normalized in order to be consistent with the MACC data.  
28 Daily dust/total AOD<sub>550</sub> ratios ( $f_d$ ) from the common GOCART-MACC period 2003-2007 were  
29 first brought to a common 1° x 1° spatial resolution using bilinear interpolation and then we  
30 calculated the regression line for each grid cell on a seasonal basis. The linear relations were  
31 afterwards used in order to normalize the 2000-2002 GOCART ratios to have a homogeneous  
32 dataset. The slopes and offsets of these regression lines and the corresponding correlation  
33 coefficients (R) can be seen in Figs. S5, S6 and S7 of the Supplement, respectively. Overall, for

1 the whole time period, the MACC reanalysis  $f_d$  ratios are lower by  $\sim 26\%$  from the GOCART  $f_d$   
2 ratios and the linear relation connecting the two products is  $f_{d\text{MACC}}=0.4964f_{d\text{GOCART}}+0.0952$   
3 with a correlation coefficient  $R$  of 0.74. The  $f_d$  values of the merged GOCART-MACC (2000-  
4 2012) timeseries were checked using the Standard Normalized Homogeneity Test (SNHT) as  
5 described in Alexandersson (1986). The statistical significance was checked following Khaliq  
6 and Ouarda (2007) and the  $f_d$  timeseries were found to be homogeneous (see Fig. S8 of the  
7 Supplement). Hence, this test verifies that the use of the merged GOCART-MACC  $f_d$  dataset  
8 will not insert any artifacts (e.g. trends or breaks) in the algorithm. Finally, the  $f_d$  data were  
9 brought to the same spatial resolution with MODIS data ( $0.1^\circ \times 0.1^\circ$ ) using bilinear interpolation  
10 again.

11 After the calculation of  $\tau_d$  with the use of  $f_d$  values ( $\tau_d=f_d\text{AOD}_{550}$ ), we proceed to the calculation  
12 of the anthropogenic contribution to the total  $\text{AOD}_{550}$  ( $\tau_a$ ) by multiplying the non-dust part of  
13  $\text{AOD}_{550}$  with the anthropogenic fraction  $f_a$  for the region of Eurasia ( $0.77\pm 0.20$ ) given in  
14 Bellouin et al. (2013) [ $\tau_a=f_a(1-f_d)\text{AOD}_{550}$ ]. The rest of the total  $\text{AOD}_{550}$  is attributed to the fine  
15 mode natural aerosols [ $\tau_n=(1-f_a)(1-f_d)\text{AOD}_{550}$ ] (see Fig. 2). As discussed in Bellouin et al.  
16 (2013), the fine mode natural aerosols consist of sea salt, dimethyl sulfide from land and  
17 oceanic sources,  $\text{SO}_2$  from degassing volcanoes and secondary organic aerosols from biogenic  
18 emissions. It has to be highlighted that like in the case of oceanic regions the biomass burning  
19 aerosols are classified as anthropogenic by this algorithm. As shown in Figs. S4a-d, the seasonal  
20 variability of  $\tau_d$  over land covered regions is very close to the LIVAS dust  $\text{AOD}_{532}$  which is  
21 used as a reference.

22 Overall, the algorithm described above performs well as far as dust is concerned. This is further  
23 shown when comparing MODIS Terra and Aqua  $\tau_d$  values with collocated AERONET  
24 observations for dust dominated days (see Fig. S9 in the Supplement). The method followed for  
25 the collocation of the data is similar to the one presented in Sect. 4.1 while dust dominated days  
26 were days with an AERONET AE smaller than 1 (see Mateos et al., 2014) and a MODIS based  
27  $\tau_d$  greater than  $\tau_a$  and  $\tau_n$  or  $\tau_m$ . The uncertainties of the calculated  $\tau_a$ ,  $\tau_d$ ,  $\tau_n$  and  $\tau_m$  values which  
28 are inserted by the input data and the assumptions of the algorithm are expected to be similar  
29 with the ones presented in Bellouin et al. (2013). Bellouin et al. (2013) using a Monte-Carlo  
30 analysis indicated that  $\tau_a$  can be specified with an uncertainty of  $\sim 23\%$  over land and  $\sim 16\%$   
31 over the ocean,  $\tau_d$  can be specified with an uncertainty of  $\sim 19\%$  over land and  $\sim 33\%$  over the  
32 ocean,  $\tau_n$  can be specified with an uncertainty of  $\sim 41\%$  and  $\tau_m$  with an uncertainty of  $\sim 28\%$ .  
33 The results of the application of the algorithm described in the paragraphs above are presented



1 in the following section (Sect. 4) by means of maps, pie charts, plots and tables for each one of  
2 the three basic regions of interest and their nine sub-regions.

## 3 4 **4 Results and discussion**

### 5 **4.1 Validation of MODIS gridded data using ground-based observations**

6 As discussed in Sects. 2 and 3, the high quality (QAC: 3) DT level-2 Collection 051 MODIS  
7 data used in this work were validated in detail against data from 13 AERONET stations (see  
8 Fig. 1). The stations were selected to make sure that their version 2.0 level 2.0 high quality  
9 cloud screened Cimel sunphotometric observations were covering at least 2 years and there  
10 were at least 100 common days of AERONET and MODIS observations. The exact  
11 geolocation of the AERONET stations is given in Table 1 (see also Fig.1) along with the  
12 period of available data, the hosting country, the type of the station (e.g. urban/rural,  
13 coastal/continental, etc.) and the corresponding mean overpass time of Terra and Aqua  
14 MODIS. First, we collocated spatially and temporally the MODIS and AERONET  
15 observations by temporally averaging AERONET measurements within  $\pm 30$  min from the  
16 MODIS overpass time (see Levy et al., 2010) and spatially averaging MODIS measurements  
17 centered within a  $25 \times 25 \text{ km}^2$  window around each station (see Koukouli et al., 2010). The  
18 use of a collocation window equal to the one used for the gridding procedure, practically,  
19 allows us to validate at the same time the  $0.1^\circ \times 0.1^\circ$  MODIS gridded product.

20 The regression lines between MODIS and AERONET AODs are shown in Fig. 3 while  
21 details about the validation results can be found in Table 2. Overall, the MODIS Terra DT  
22 Collection 051 data overestimate  $\text{AOD}_{550}$  by 11.59 % (Normalized Mean Bias - NMB) with  
23 63.28 % of the data falling within the expected error (EE) envelope and 67.78% within the  
24 pre-launch expected error (pLEE) envelope. The expected error envelope is define as:  $\text{AOD} -$   
25  $|\text{EE}| \leq \text{AOD}_{\text{MODIS}} \leq \text{AOD} + |\text{EE}|$  with EE being  $\pm(0.05+0.15\text{AOD})$  (Levy et al., 2010) and  
26 pLEE being  $\pm(0.05+0.20\text{AOD})$  (Kaufman et al., 1997). On the other hand, the MODIS Aqua  
27 DT Collection 051 data overestimate  $\text{AOD}_{550}$  by 25.18 % (NMB) with 57.14 % of the data  
28 falling within the EE envelope and 61.87 % within the pLEE envelope. The percentage of the  
29 MODIS Terra and Aqua data falling within the EE envelope are close to the 57 % given in  
30 Remer et al. (2005) for the Eastern Mediterranean. The validation results for each station  
31 separately can be found in Table S1 of the Supplement. The results discussed in this  
32 paragraph are comparable to the ones appearing in previous studies focusing on the  
33 Mediterranean region (see Papadimas et al., 2009; Koukouli et al., 2010). In general, it is

1 shown here that the MODIS Terra Collection 051 data exhibit a better agreement with the  
2 ground-based observations from AERONET than MODIS Aqua data do. Therefore, the  
3 statistics appearing for MODIS Terra throughout the paper could be considered more robust.  
4 To be in line with the global validation of the DT Collection 051 product by Levy et al.  
5 (2010) we also performed a validation with the specifications used in their work. We used a  
6  $50 \times 50 \text{ km}^2$  window for the spatial collocation of the MODIS and AERONET data while  
7 only days with at least 5 MODIS retrievals and 2 AERONET measurements were taken into  
8 account. The increased size of the collocation window improves the results of the validation.  
9 As shown in Table 2, MODIS Terra DT Collection 051 data overestimate  $\text{AOD}_{550}$  by 5.10 %  
10 (NMB) with 70.17 % of the data falling within the EE envelope and 74.64 % within the pLEE  
11 envelope. For MODIS Aqua, the NMB is 15.34%, while the percentage of the measurements  
12 falling within the EE and pLEE envelope is 66.76 % and 70.45 %, respectively. These results  
13 for the Eastern Mediterranean are close to the global ones presented in Levy et al. (2010).  
14 As discussed in Sect. 3.1, data from the DB algorithm were used over bright arid and semi-  
15 arid regions of North Africa for the production of the  $0.1^\circ \times 0.1^\circ$  MODIS gridded dataset for  
16 grid cells with no DT data. Therefore, in this work we also perform a validation of the DB  
17 Collection 051 product over the region of the Eastern Mediterranean. In the case of DB data,  
18 we first make use of all the available DB observations without any quality filtering over the  
19 13 AERONET stations. A spatial window of 25-30 km has been typically used in the past for  
20 the collocation of MODIS DB data with the AERONET observations (see Shi et al., 2011;  
21 Ginoux et al., 2012; Sayer et al., 2013; 2014) which is in line with the  $25 \times 25 \text{ km}^2$  window  
22 used here. The MODIS Terra DB data overestimate  $\text{AOD}_{550}$  by 21.38 % (NMB) with 51.90 %  
23 of the data falling within the expected uncertainty (EU) envelope assuming a DB expected  
24 uncertainty of  $\pm 0.05 \pm 20\% \text{AOD}_{\text{AERONET}}$  (Hsu et al., 2006). The MODIS Aqua DB Collection  
25 051 data overestimate  $\text{AOD}_{550}$  by 33.03 % (NMB) with 55.30 % of the data falling within the  
26 expected uncertainty envelope. We repeated the validation procedure for DB data taking into  
27 account the highest quality data only. The sample of available measurements was diminished  
28 by a factor of 5 in the case of MODIS Terra and 6 in the case of MODIS Aqua but the results  
29 were pretty similar with the ones for the unfiltered data. Therefore, the use of unfiltered DB  
30 data during the gridding procedure does not insert any significant uncertainty. The DB results  
31 for the 13 AERONET stations examined here are not of the same agreement with the DT  
32 results and the ones presented in previous studies utilizing DB Collection 051 data for other  
33 stations and larger regions (see Shi et al., 2011; Ginoux et al., 2012). However, it has been

1 reported that stations in the region (e.g. Sede Boqer in Israel) are among the ones with the  
2 greatest discrepancies between MODIS DB and AERONET measurements (Ginoux et al.,  
3 2012). Nevertheless, as commented in Sect. 3.1, the DB data constitute only a small fraction  
4 of the data used for the production of the MODIS gridded dataset (~ 1 % only of the 30000  
5 grid cells covering the Eastern Mediterranean has only DB retrievals) and therefore they do  
6 not affect significantly its quality. Only areas in Northern Africa are expected to be affected  
7 by the use of DB data due the extended lack of DT data there.

8 As discussed in Sect. 3.1 the gridding procedure was repeated four times using a gridding  
9 window of 25, 50, 75 and 100-km using MODIS Terra AOD<sub>550</sub> data for the year 2004 showing  
10 that the 25-km window is optimal for capturing local pollution sources. In order to see how the  
11 size of the gridding window affects the agreement between MODIS and AERONET data we  
12 also proceeded to a validation of MODIS DT data against AERONET measurements using  
13 different spatial collocation windows (25, 50, 75 and 100-km) and two quality criteria, a "strict"  
14 one: at least 2 AERONET measurements for each MODIS-AERONET pair and a "stricter" one:  
15 at least 5 MODIS retrievals and 2 AERONET measurements for each MODIS-AERONET pair  
16 as in Levy et al. (2010). The results for the DT MODIS Terra and Aqua data are presented in  
17 Table S2 of the Supplement. In general, it is shown that the increased size of the spatial  
18 collocation window leads to an improvement of the bias between satellite and ground-based  
19 observations. This is probably due to the inclusion of more observations into the calculations  
20 which diminishes the noise of the MODIS observations. In addition, as expected, the stricter  
21 quality criteria lead to a better agreement between MODIS DT and AERONET data. Taking  
22 into account not only the NMB but also the regression lines and the other metrics appearing in  
23 Table 2S, it is concluded that the 50-km window is the best choice for the validation procedure  
24 in line with Ichoku et al. (2002). On the other hand, the 25-km validation results are close to the  
25 50-km ones (see Table S2) and at the same time the 25-km gridding window allows for a more  
26 efficient detection of local aerosol sources as shown in Sect. 3.1. Taking this into account, we  
27 suggest that the 25-km window used for the production of the 0.1° x 0.1° gridded MODIS  
28 dataset is the optimal selection for studying the spatial variability of aerosols, preserving at the  
29 same time the representativeness of the real aerosol load over each specific spot.

30

## 31 **4.2 Aerosol spatial variability and hot spots**

32 The AOD<sub>550</sub> spatial variability over the greater Eastern Mediterranean region for the period  
33 2000-2012 as seen from the Terra MODIS 0.1° x 0.1° dataset is presented in Fig. 4. Several

1 aerosol hot spots that coincide with megacities (e.g. Cairo, Istanbul), large cities (e.g. Athens,  
2 Ankara, Alexandria, Izmir, Thessaloniki) or even medium sized cities (e.g. Larissa,  
3 Limassol), industrial zones (e.g. OSTIM Industrial Zone in Ankara, Turkey), power plant  
4 complexes (e.g. Maritsa Iztok complex at the Stara Zagora Province in Bulgaria, Ptolemaida-  
5 Kozani power plants in Western Macedonia, Greece), river basins (e.g. Evros river Basin at  
6 the borders between Greece and Turkey), etc, can be detected on the map. Indicatively, in Fig  
7 4 we give a list of 35 local particle pollution sources in the region; however, careful  
8 inspection of this map and the seasonal maps presented in Fig. 6 allows for the detection of  
9 many more aerosol sources. The results from the analysis of Aqua MODIS data are pretty  
10 similar as shown in Fig. S10 of the Supplement. A significant number of the local aerosol  
11 sources can also be detected on the OMI 2004-2012 tropospheric NO<sub>2</sub> and PBL SO<sub>2</sub> maps  
12 given in Figs. 5a and b which reveals the origin of aerosols over these regions (e.g. traffic,  
13 industrial activities, etc). However, there are regions of high aerosol load which cannot be  
14 seen in Fig. 5a and b and vice versa which is indicative of the significant role of other  
15 anthropogenic or natural processes that contribute to the local aerosol load (e.g. fires, soil dust  
16 from agricultural activities or arid regions, Sahara dust transport).

17 The topography (Fig. 5c) and precipitation (see Fig. 5d for annual precipitation levels for the  
18 period 2000-2012 from TRMM) are also major determinants of the local AOD<sub>550</sub> levels. For  
19 example, regions with mountain ranges in the Balkan Peninsula (e.g. Pindus mountain range  
20 in Greece, Dinaric Alps that run through Albania and the former Yugoslav republics, the  
21 Balkan mountain range in Central Bulgaria) are characterized by low AODs (see Fig. 4). On  
22 the contrary, regions of low altitude are generally characterized by higher AODs because the  
23 majority of anthropogenic activities is usually concentrated there. Also, low altitude regions  
24 surrounded by high mountains are characterized by higher AODs as aerosols cannot be easily  
25 transported by the wind (e.g. the industrialized regions in Central Bulgaria which are confined  
26 between the high Balkan and Rodopi mountain ranges). As precipitation is the major removal  
27 mechanism of pollutants in the atmosphere, regions with high AOD<sub>550</sub> are in many cases  
28 connected to low precipitation levels and vice versa (see Figs. 4 and 5d). It has to be  
29 highlighted that the AOD<sub>550</sub> over these regions is high primarily due to the emissions and the  
30 atmospheric processes forming aerosol particles. The low removal rates from precipitation  
31 just preserve the AOD<sub>550</sub> levels high. A striking example is the region of Anatolia in Central  
32 Turkey which is characterized by lower precipitation levels and higher aerosol loads

1 compared to the surrounding regions. Also, the low precipitation levels are partly responsible  
2 for the high aerosol loads appearing over Northern Africa.

3 Overall, the mean AOD<sub>550</sub> for the whole period of interest is estimated at  $0.215 \pm 0.187$  for  
4 Terra and  $0.217 \pm 0.199$  for Aqua MODIS for the Eastern Mediterranean region which is ~ 45  
5 % higher than the global average appearing in recent studies (e.g. Kourtidis et al., 2015). Over  
6 land higher mean AODs are generally recorded ( $0.219 \pm 0.165$  for Terra and  $0.239 \pm 0.189$   
7 for Aqua MODIS) than over the sea ( $0.213 \pm 0.201$  for Terra and  $0.202 \pm 0.205$  for Aqua  
8 MODIS). All these values along with the mean AODs for the 9 sub-regions of interest  
9 covering the Eastern Mediterranean can be found in Table 3.

10 The AOD<sub>550</sub> spatial variability on a seasonal basis from MODIS Terra and Aqua is presented  
11 in Fig. 6 along with the difference between the two products. The majority of the local aerosol  
12 sources over land are more prominent in summer. The limited washout by precipitation (see  
13 also Papadimas et al., 2008) and also the enhanced photochemical production of secondary  
14 organic aerosols (Kanakidou et al., 2011 and references therein) contribute to the high AODs  
15 appearing over local sources. In addition, during summer, over the region, there is typically a  
16 significant transport of aerosols (e.g. see Kanakidou et al., 2011 and references therein) and  
17 gaseous pollutants like SO<sub>2</sub> and NO<sub>2</sub> (see Georgoulas et al., 2009; Zyrichidou et al., 2009)  
18 and biomass burning aerosols from Central-Eastern Europe. Over the sea, a profound  
19 maximum is observed in spring extending across the North African coast and the neighboring  
20 oceanic areas which is due to the well documented transport of significant amounts of dust  
21 from the Sahara Desert (see Barnaba and Gobbi., 2004 and the list of references given in the  
22 introduction). The seasonal variability of aerosols and the relative role of different aerosol  
23 types and various processes is discussed in more details in Sect 4.4.

24 The difference between MODIS Terra and Aqua Collection 051 AOD<sub>550</sub> over the Eastern  
25 Mediterranean is -0.002 (-1.40 %) for winter, -0.009 (-3.27 %) for spring, -0.011 (-4.46 %) for  
26 summer and 0.008 (4.40 %) for autumn. AOD<sub>550</sub> levels from Terra MODIS are lower than  
27 that from Aqua MODIS over land for all seasons. Over the sea, Terra MODIS AOD<sub>550</sub> levels  
28 are lower than that of Aqua MODIS only in winter. The fact that Terra MODIS measurements  
29 are systematically higher than that from Aqua over the sea by ~ 0.01 on an annual basis is in  
30 line with the findings of previous global studies for Collection 5 (e.g. Remer et al., 2006;  
31 2008). Locally, one can see regions with positive and negative differences between Terra and  
32 Aqua MODIS AOD<sub>550</sub>. The patterns of the Terra-Aqua difference per season are presented in  
33 Figs. 6c, f, i and l while the patterns of the percent difference are given in Fig. S11 of the

1 Supplement. The largest part of the Terra-Aqua MODIS differences over land and sea which  
2 are observed here may be attributed to the known calibration and sensor degradation issues of  
3 MODIS (for details see Levy et al., 2010; 2013; Lyapustin et al., 2014; Georgoulias et al.,  
4 2016). A significant effort has been undertaken to address these issues in the new (Collection  
5 6) MODIS product (e.g. Levy et al., 2013; Lyapustin et al., 2014; Georgoulias et al., 2016)  
6 and a repetition of a similar analysis with Collection 6 data in the future would be a valuable  
7 contribution. Taking into account the aforementioned issues and the retrieval uncertainty of  
8 MODIS it becomes more than obvious that the attribution of observed differences between  
9 Terra and Aqua to the diurnal variability of aerosol load (e.g. over biomass burning regions)  
10 in the region is a difficult task. It is shown in Fig. S12 of the Supplement that the diurnal  
11 variability of AOD<sub>550</sub> from AERONET ranges significantly from station to station. The  
12 average hourly departure from the daily mean for the total of the 13 stations ranges from ~ -5  
13 % to ~ 5 %. Specifically, for the MODIS Terra and Aqua overpass times, the AERONET  
14 AOD<sub>550</sub> difference ranges from ~ -10 % to ~ 10 % (see Fig. S12b). The Terra-Aqua AOD<sub>550</sub>  
15 difference is negative for the total of the 13 stations ranging from ~ -25 % to ~ -5 %. It is  
16 shown in Fig. S12b that the two differences exhibit a similar variability from station to station  
17 which indicates that part of the observed Terra-Aqua difference is indeed due to the diurnal  
18 variability of aerosols. However, as mentioned above, the diurnal variability of aerosols is a  
19 very delicate issue and should be comprehensively addressed in a future study. The same  
20 stands for other kind of variabilities which could be connected to local and regional  
21 anthropogenic activities like e.g. the weekly cycle of aerosols (see Georgoulias and Kourtidis,  
22 2011; Georgoulias et al., 2015).

23

### 24 **4.3 Contribution of different aerosol types to the total AOD<sub>550</sub>**

#### 25 **4.3.1 Annual contribution**

26 As mentioned above, we attempt to estimate in our work the contribution of different aerosol  
27 types to the total AOD<sub>550</sub> over the region of the Eastern Mediterranean was calculated  
28 following the methodology presented in Sect. 3.2. For the land covered areas, based on  
29 MODIS Terra observations, we estimate that 52 % (0.112±0.087) of the total AOD<sub>550</sub> is due  
30 to anthropogenic aerosols, 32 % (0.074±0.080) due to dust and 16 % (0.034±0.026) due to  
31 fine mode natural aerosols (see Fig. 7). For the oceanic areas, 41 % (0.086±0.085) of the total  
32 AOD<sub>550</sub> is due to anthropogenic aerosols, 34 % (0.076±0.185) due to dust and 25 %  
33 (0.054±0.018) due to marine aerosols (see Fig. 7). The results based on observations from

1 MODIS Aqua are similar. Over land, 50 % ( $0.117\pm 0.093$ ) of the total  $AOD_{550}$  is  
2 anthropogenic, 35 % ( $0.090\pm 0.102$ ) is due to dust and 15 % ( $0.035\pm 0.028$ ) due to fine mode  
3 natural aerosols, while, over the sea, 40 % ( $0.079\pm 0.080$ ) of the total  $AOD_{550}$  is of  
4 anthropogenic origin, 33 % ( $0.070\pm 0.181$ ) is due to dust and 27 % ( $0.054\pm 0.018$ ) due to  
5 marine aerosols (see Fig. 7). These results along with the relative contributions and the annual  
6  $\tau_a$ ,  $\tau_d$ ,  $\tau_n$  and  $\tau_m$  levels for each one of the nine sub-regions of interest (see Fig. 1) are given in  
7 Table 4.

8 For anthropogenic aerosols, the region with the highest relative contribution is NBL (59 % for  
9 both Terra and Aqua MODIS) while the region with the lowest relative contribution is SWO  
10 (32 % for both Terra and Aqua MODIS) (see also Table 4). The spatial variability of  $\tau_a$  is  
11 presented in Fig. 8a for Terra MODIS and Fig. S13a of the Supplement for Aqua MODIS, the  
12 patterns being similar in both cases. Over land, the annual  $\tau_a$  patterns are similar to the  
13  $AOD_{550}$  patterns, the highest values appearing over local particle pollution sources (cities,  
14 industrial zones, etc.). Over the sea,  $\tau_a$  is higher along the coasts, while it drops significantly  
15 towards other directions. An interesting feature here is that the oceanic region of Black Sea  
16 (BSO) presents higher relative anthropogenic contributions than the rest of the oceanic sub-  
17 regions but also than land areas with significant anthropogenic sources (e.g. ANL and NAL).  
18 This is indicative of the transport of atmospheric particles from Central Europe and biomass  
19 burning aerosols during the biomass burning seasons in April-May from Russia (across the  
20 latitudinal zone  $45^\circ\text{N}$ - $55^\circ\text{N}$ ) and July-August from South-Western Russia and Eastern Europe  
21 (Amiridis et al., 2010). These aerosols are transported at much lower latitudes as shown in  
22 previous studies (e.g. Vrekoussis et al., 2005; Karnieli et al., 2009) reaching the Sahara Desert  
23 and the Middle East regions (Pozzer et al., 2015). The fact that  $\tau_a$  drops gradually from the  
24 coasts is also seen in Fig. 9 where the latitudinal variability of the optical depths of the  
25 different aerosol types ( $\tau_a$ ,  $\tau_d$ ,  $\tau_n$  and  $\tau_m$ ) is presented for four bands that cover the whole  
26 Eastern Mediterranean. An interesting feature is that  $\tau_a$  increases nearby the shoreline  
27 (particularly along the North African coastal zone) before it gradually decreases. Over land  
28 aerosols are located within the atmospheric boundary layer, close to the emission sources, and  
29 hence, their deposition and removal from the atmosphere is more efficient than over the sea.  
30 The particles which are transported over the sea on the other hand usually reach greater  
31 heights which prolongs their lifetime.

32 As shown in Fig. 9, the same feature is observed for dust. Indicatively,  $\tau_d$  and the relative  
33 contribution of dust to the total  $AOD_{550}$  on an annual basis over the oceanic regions of SWO

1 and SEO are in general higher or comparable to the ones over NAL (see Table 4 for more  
2 details). In Fig. 9, the MODIS-based  $\tau_d$  latitudinal variability is presented along with the  
3 latitudinal variability of dust AOD<sub>532</sub> and extinction coefficients of dust at 532 nm from  
4 LIVAS. As expected, in all cases  $\tau_d$  decreases with distance from the large dust sources in the  
5 South and South-East (Sahara Desert, Middle East deserts) with local maxima over the  
6 latitudinal zone from 35°N to 40°N (especially for band 2 and band 3). The latitudinal  
7 variability of  $\tau_d$  is similar to the latitudinal variability of dust AOD<sub>532</sub> for all the four bands  
8 despite the fact that the MODIS-based data have a resolution 100 times higher (0.1° vs 1°) and  
9 therefore are more sensitive to local characteristics. Dust reaches heights up to ~ 4-5 km in the  
10 area; however, the largest fraction of dust mass is confined within the first 2-3 km of the  
11 troposphere (see Fig. 9). The annual  $\tau_d$  patterns are shown in Fig. 8b for Terra MODIS (Fig.  
12 S13b of the Supplement for Aqua MODIS). The main dust transport pathways over the  
13 oceanic areas of the Eastern Mediterranean can be seen along with various local maxima over  
14 land. The highest  $\tau_d$  values over land appear over the regions of NAL and ANL (see Table 4)  
15 and along the coasts. The high dust concentrations appearing over these regions are not only  
16 due to the transport of dust from the nearby deserts but also due to local dust sources. A  
17 recent study by Liora et al. (2015) reports various local sources of wind blown dust along the  
18 coastal regions of Greece and Turkey, over the region of Anatolia in Turkey, over the Greek  
19 islands, Crete, Cyprus and regions close to the coastal zone of Middle East. Their results are  
20 in good agreement with the  $\tau_d$  patterns presented in this work.

21 As shown in Fig. 7, fine mode natural aerosols exhibit the lowest contribution to the total  
22 AOD<sub>550</sub> compared to the other aerosol types over land. The spatial variability of  $\tau_n$  is very low  
23 compared to  $\tau_a$  and  $\tau_d$  as shown in Figs. 8c and 9. It is inferred from the values appearing in  
24 Table 4 that  $\tau_n$  increases slightly as one moves from North to South; however, the relative  
25 contribution of fine mode natural aerosols to the total AOD<sub>550</sub> slightly decreases (i.e. 17.67 %  
26 over NBL and 14.97 % over NAL according to Terra MODIS observations). The latitudinal  
27 variability and the percentages appearing in Table 4 are in accordance to the relative  
28 contributions of biogenic aerosols to the total AOD<sub>550</sub> appearing over the Eastern  
29 Mediterranean in a recent modeling study (Rea et., 2015).

30 Similar to fine mode natural aerosols over land, marine aerosols generally have the lowest  
31 contribution to the total AOD<sub>550</sub> compared to the other aerosol types over the sea (see Fig. 7  
32 and Table 4) except for BSO. The variability of  $\tau_m$  is very low compared to  $\tau_a$  and  $\tau_d$ . On an  
33 annual basis, high  $\tau_m$  values appear over the Aegean Sea and the oceanic area between Crete



1 and the North African coast while slightly lower values appear along the coasts of the Eastern  
2 Mediterranean (see Figs. 8d and 9). The  $\tau_m$  patterns follow the near surface wind speed  
3 patterns in the region (see Fig. S14 of the Supplement) being in accordance to the  $\tau_m$ , marine  
4 particulate matter concentration or sea salt emission patterns appearing in other studies (Im et  
5 al., 2012; Nabat et al., 2013; Rea et al., 2015; Liora et al., 2015).

#### 6 7 **4.3.2 Seasonal contribution**

8 The contribution of different aerosol types to the total AOD<sub>550</sub> over the Eastern Mediterranean  
9 varies from season to season. The relative contribution of each aerosol type over EML and  
10 EMO for each season is shown in Fig. 10. Over land, the relative contribution of  $\tau_a$ ,  $\tau_d$  and  $\tau_n$   
11 to the total AOD<sub>550</sub> exhibits a low seasonal variability. The relative contribution of  
12 anthropogenic aerosols to the total AOD<sub>550</sub> ranges from 49 % in SON to 55 % in DJF based  
13 on Terra MODIS observations and from 48 % in MAM and SON to 52 % in JJA based on  
14 Aqua MODIS observations. In contrast, over the oceanic regions the relative contribution of  
15  $\tau_a$ ,  $\tau_d$  and  $\tau_m$  to the total AOD<sub>550</sub> exhibits a significant seasonal variability. The relative  
16 contribution of anthropogenic aerosols to the total AOD<sub>550</sub> ranges from 27 % / 27 % in DJF to  
17 50 % / 47 % in JJA based on Terra/Aqua MODIS observations. The percentages appearing  
18 here are in accordance to the values appearing in Hatzianastassiou et al. (2009) where a  
19 different satellite-based approach was followed. Indicatively, for the greater Athens area, an  
20 average summertime anthropogenic contribution of ~ 50 % was found here based on Terra  
21 MODIS data which is within the summer period range of 47-61 % indicated in the study by  
22 Hatzianastassiou et al. (2009). In addition, the corresponding values for the greater  
23 Thessaloniki area, Crete, Cairo and Alexandria are 53 %, 38 %, 48 % and 41 %, respectively,  
24 within the range of values (57-73 %, 36-52 %, 34-56 % and 23-60 %) shown in  
25 Hatzianastassiou et al. (2009). Only in the case of Ankara, our results suggest a lower  
26 anthropogenic contribution (52 % versus 71-84 %). Particularly for Athens, Gerasopoulos et  
27 al. (2011) following a different approach incorporating ground-based AOD observations and  
28 trajectory modeling reached similar results (annual contribution of ~ 62 % from local and  
29 regional sources and continental Europe which is expected to be mostly of anthropogenic  
30 origin). Similarly, for Crete, Bergamo et al. (2008) using a different approach, also utilizing  
31 ground-based data, found an annual anthropogenic contribution of ~ 43 %.

32 The seasonal patterns of the anthropogenic aerosols ( $\tau_a$ ) over the Eastern Mediterranean based  
33 on MODIS Terra observations are presented in Figs. 11a, e, i and m while the seasonal

1 variability of  $\tau_a$  over the whole region, over the land covered part and the oceanic part and  
2 over the 9 sub-regions of interest is presented in Fig. 12. The results based on MODIS Aqua  
3 observations are similar and can be found in Figs. S15a, e, i and m and Fig. S16 of the  
4 Supplement. Generally, the local hot spots are detectable throughout the year; however, they  
5 are becoming much more discernible in spring and especially in summer. As shown in Fig.  
6 12a,  $\tau_a$  nearly doubles during the warm period of the year (spring-summer) with the seasonal  
7 variability being stronger over the sea (Fig. 12c) than over land (Fig. 12b). A clear peak is  
8 observed in summer, August being the month with highest  $\tau_a$  levels. As discussed in Sect.  
9 4.3.1 the summer peak is mostly a result of three basic reasons. The first one is the deficiency  
10 of wet removal processes compared to the cold period. As shown in Fig. S17, based on the  
11 TRMM satellite observations, August and July are the months with the lowest precipitation  
12 levels over the land covered part (a drop of  $\sim 75\%$  compared to winter months) and over the  
13 oceanic part (a drop of  $\sim 90\%$  compared to winter months) of the Eastern Mediterranean,  
14 respectively. The second reason is the enhancement of the photochemical production of  
15 secondary organic aerosols in summer (Kanakidou et al., 2011) and the third reason is the  
16 transport of pollution aerosols from Central Europe and biomass burning aerosols from South-  
17 Western Russia and Eastern Europe during the biomass burning season in July-August  
18 (Amiridis et al., 2010). The Etesians, which are persistent northerly winds that prevail over  
19 the Eastern Mediterranean during summer, bring dry and cool air masses and aerosols from  
20 the regions mentioned above while blocking at the same time the transport of desert dust in  
21 the region and dispersing local pollution in urban areas to levels typical for rural areas (see  
22 Tyrlis and Lelieveld, 2013 and references therein). As seen in Figs. 12a-l, a smaller but  
23 distinct in most cases  $\tau_a$  peak appears in April mostly as a result of the transport of biomass  
24 burning aerosols from Russia (across the latitudinal zone  $45^\circ\text{N}$ - $55^\circ\text{N}$ ). This is in line with the  
25 findings of Sciare et al. (2008) who detected traces of these biomass burning aerosols at the  
26 island of Crete in Southern Greece.

27 As discussed above, the relative contribution of dust to the total  $\text{AOD}_{550}$  over land exhibits a  
28 low seasonal variability ranging from 29 % in DJF to 36 % in SON based on Terra MODIS  
29 observations and from 33 % in JJA to 38 % in SON based on Aqua MODIS observations (see  
30 Fig. 10). Over the oceanic regions the relative contribution of dust to the total  $\text{AOD}_{550}$  ranges  
31 significantly throughout a year from 26 % / 28 % in JJA to 42 % / 39 % in MAM based on  
32 Terra/Aqua MODIS observations. The percentages appearing here are in accordance to model  
33 and observational studies. For example, de Meij et al. (2012) using the atmospheric chemistry

1 general circulation model EMAC (ECHAM/MESSy Atmospheric Chemistry) showed that  
2 dust contributes on an annual level  $\sim 30\%$  to the total  $AOD_{550}$  over stations located in the  
3 area of the Eastern Mediterranean. Gerasopoulos et al. (2011) found a  $\sim 23\%$  percent  
4 contribution of North African dust to the total AOD over Athens using ground-based AOD  
5 observations and trajectory modeling. Taking into account that part of the  $\sim 39\%$  local and  
6 regional sources appearing in Gerasopoulos et al. (2011) is due to local dust sources,  
7 especially in summer, turns out that their results are in agreement with the  $\sim 33\%$  relative  
8 contribution found in this work for the greater Athens area based on Terra MODIS  
9 observations. The seasonal patterns of dust ( $\tau_d$ ) over the region based on Terra MODIS  
10 observations are shown in Figs. 11b, f, j and n while the seasonal variability of  $\tau_d$  over the  
11 whole region, over land, over the sea and over the 9 sub-regions of interest is shown in Fig.  
12 12. The corresponding results based on MODIS Aqua observations are pretty similar and can  
13 be found in Figs. S15b, f, j and n and Fig. S16 of the Supplement.

14 As seen in Fig. 11f, in spring, mostly due to the strong Sahara dust events, very high  $\tau_d$  values  
15 appear over land regions in North Africa, Middle East, Anatolia and oceanic areas across the  
16 Eastern Mediterranean (especially below  $35^\circ\text{N}$ ). Dust loading over the sea exhibits two  
17 maxima, one at the coastal zone of Libya and one across the coastal zone of Middle East. The  
18 same two maxima but with much lower  $\tau_d$  values appear in summer (Fig. 11j) and autumn  
19 (Fig. 11n). Over land, the  $\tau_d$  patterns are similar in summer and autumn, the maximum values  
20 appearing over the Anatolian Plateau and areas of North Africa and Middle East. During  
21 winter, dust maxima appear across the coastal zone of Northern Africa with relatively low  $\tau_d$   
22 values across the coastal zone of Middle East (Fig. 11b). In winter  $\tau_d$  levels are low over land  
23 compared to the other seasons (Figs. 11b, f, j and n) as precipitation levels (see Fig. S17 of the  
24 Supplement) and hence wet scavenging of aerosols peak. At the same time, the local  
25 emissions of dust are low for regions away from the large area sources in the South (Liora et  
26 al., 2015). In contrast, over the sea  $\tau_d$  levels in winter are similar or slightly higher for some  
27 areas than that in summer and autumn (see Figs. 11 and 12) as this is the season with the  
28 second highest frequency (after spring) of strong ( $\sim 21\%$ ) and extreme ( $\sim 26\%$ ) desert dust  
29 episodes in the region (see Gkikas et al., 2013 for details). February is by far the winter month  
30 with the highest  $\tau_d$  levels (see Fig. 12) in line with the findings of Pey et al. (2013) who  
31 showed that the intensity of African dust episodes over stations in Greece and Cyprus peaks  
32 in February. Dust exhibits a strong peak in spring, April being the month with the highest  $\tau_d$   
33 levels in line with other studies (e.g. Israelevic et al., 2012; Varga et al., 2014). The peak in

1 April is a result of the high cyclonic activity over North Africa during this month as shown by  
2 Flaounas et al. (2015). According to the same study, low pressure systems are responsible for  
3 ~ 10-20 % of moderate and ~ 40-50 % of high and extreme Sahara dust transport events over  
4 the Eastern Mediterranean. North Africa (Sharav) cyclones develop mainly in spring and  
5 summer while Mediterranean cyclones develop in winter and autumn. The Mediterranean  
6 cyclones are more intense than Sharav cyclones. The region, is also affected by events  
7 bringing particles from dust source regions in the eastern part of the Mediterranean basin  
8 (Negev desert in Israel, Sinai in Egypt, Anatolian Plateau in Turkey) and the Arabian deserts  
9 (Basart et al. 2009; Pey et al., 2013; Abdelkader et al., 2015). Dust remains in the atmosphere  
10 for a period of 1-4 days undergoing chemical aging before being removed (see Abdelkader et  
11 al. 2015 and references therein). The seasonal variability of  $\tau_d$  is much stronger and the spring  
12 maxima much more prominent over the sea (see Fig. 12). This is expected, as dust is only  
13 occasionally transported over the sea during episodic events, while over land, local sources  
14 also contribute to the dust burden especially in summer due to the dryness of soil. For  
15 example, over NBL, a broad spring-summer peak is observed, June being the month with the  
16 highest  $\tau_d$  levels. As one moves south (SBL, ANL and NAL) the April peak becomes more  
17 prominent.

18 The relative contribution of fine mode natural aerosols to the total  $AOD_{550}$  over land exhibits  
19 a very low seasonal variability ranging from 15 % in MAM and SON to 16 % in DJF and JJA  
20 based on Terra MODIS observations and from 14 % in DJF and SON to 15 % in MAM and  
21 JJA based on Aqua MODIS observations (see Fig. 10). The seasonal variability is also very  
22 low, the highest values appearing in spring and summer (Fig. 12). Despite the generally low  
23 contribution of fine mode natural aerosols to the total  $AOD_{550}$  over the Eastern  
24 Mediterranean,  $\tau_n$  levels are similar to  $\tau_d$  levels during winter months over specific regions  
25 (NBL and SBL). The low seasonal variability can also be seen in Figs. 11c, g, k and o where  
26 the patterns of fine mode natural aerosols ( $\tau_n$ ) are presented.

27 The seasonal relative contribution of marine aerosols to the total  $AOD_{550}$  over the oceanic  
28 regions of the Eastern Mediterranean is shown in Fig. 10.  $\tau_m$  ranges from 20 % in MAM to 35  
29 % in DJF based on Terra MODIS observations and from 21 % in MAM to 36 % in DJF based  
30 on Aqua MODIS observations (see Fig. 10). Like in the case of fine mode natural aerosols,  
31 the seasonal variability is very low, but here the highest values appear in winter (Fig. 12). Due  
32 to the linear relation of  $\tau_m$  and near surface wind speed within our algorithm (see Fig. 2) the  
33  $\tau_m$  seasonal variability and patterns follow the wind speed ones (see Figs. 11d, h, l, p and

1 S14). Marine aerosol concentrations are lower close to the coastlines while the highest  
2 concentrations (see Liora et al., 2015) and  $\tau_m$  values within the Eastern Mediterranean appear  
3 over the Aegean Sea (see Fig. 11). Overall, the  $\tau_m$  patterns are in accordance to the  $\tau_m$ , marine  
4 particulate matter concentration and sea salt emission patterns from previous studies (Im et  
5 al., 2012; Nabat et al., 2013; Rea et al., 2015; Liora et al., 2015).

## 6 7 **5 Summary and conclusions**

8 In this work, satellite data from MODIS Terra (3/2000-12/2012) and Aqua (7/2002-12/2012)  
9 were analyzed separately in order to examine the spatial and temporal variability of aerosols  
10 over the Eastern Mediterranean. A high resolution ( $0.1^\circ \times 0.1^\circ$ ) MODIS gridded dataset was  
11 compiled using a method that could be used in future regional studies. A number of tests were  
12 implemented and the dataset was validated in detail using sunphotometric observations from  
13 13 AERONET stations. According to the validation, the statistics appearing for MODIS Terra  
14 throughout the paper could be considered more robust while areas in Northern Africa are  
15 expected to be affected by the extended use of DB data which do not exhibit a very good  
16 matching with the ground-based observations. It is shown that the gridding method we use  
17 offers the best compromise for studying the spatial variability of aerosols on a regional or  
18 local scale, preserving at the same time the representativeness of the real aerosol load over  
19 each specific spot.

20 Based on MODIS observations the average AOD<sub>550</sub> levels over the region of the Eastern  
21 Mediterranean are  $\sim 0.22 \pm 0.19$  which is  $\sim 45\%$  higher than the global mean. A number of  
22 aerosol hot spots that coincide with megacities, large and even medium size cities, industrial  
23 zones, power plant complexes, river basins, etc., can be detected on the AOD maps. A  
24 number of local aerosol sources can also be seen on satellite retrieved tropospheric NO<sub>2</sub> and  
25 planetary boundary layer SO<sub>2</sub> maps from OMI/AURA. This is indicative of the strong  
26 presence of anthropogenic aerosols over these regions. Topography and precipitation also  
27 play an important role. Generally, regions with mountain ranges are characterized by low  
28 AODs while regions of low altitude are characterized by higher AODs. Regions with high  
29 AOD<sub>550</sub> are in many cases connected to low precipitation levels and vice versa. Precipitation  
30 is the major washout mechanism of atmospheric pollutants. Low removal rates from  
31 precipitation contribute in preserving high the AOD<sub>550</sub> levels which are a result of emissions  
32 and other atmospheric processes.

1 The AOD<sub>550</sub> patterns over the Eastern Mediterranean exhibit a significant seasonal variability  
2 which is mostly driven by precipitation, photochemical production of secondary organic  
3 aerosols, transport of pollution and biomass burning aerosols from Central and Eastern  
4 Europe and transport of dust from the Sahara Desert and the Middle East. Differences  
5 between the MODIS Terra and Aqua Collection 051 AOD<sub>550</sub> over the Eastern Mediterranean  
6 are generally small ( $\sim -8\%$  over land and  $\sim 5\%$  over the sea). The comparison of the Terra-  
7 Aqua differences with diurnal variabilities from the AERONET stations showed that only a  
8 part of the observed differences is due to the diurnal variability of aerosols.

9 The MODIS data were combined with data from other satellites (Earth Probe TOMS,  
10 OMI/AURA), reanalysis projects (ERA-Interim, MACC) and a chemistry-aerosol-transport  
11 model (GOCART) to calculate the contribution of different types of aerosols to the total  
12 AOD<sub>550</sub>. The algorithm used was optimized for the Eastern Mediterranean through a number  
13 of tests and comparison with LIVAS CALIOP/CALIPSO dust retrievals and AERONET  
14 ground-based observations. A different approach is used for land and sea as there is not any  
15 reliable satellite retrieved quantity to separate the contribution of fine and coarse mode  
16 aerosols over water surfaces.

17 Overall, for the land areas, based on MODIS Terra observations, 52 % ( $0.112\pm 0.087$ ) of the  
18 total AOD<sub>550</sub> is due to anthropogenic aerosols, 32 % ( $0.074\pm 0.080$ ) due to dust and 16 %  
19 ( $0.034\pm 0.026$ ) due to fine mode natural aerosols (see Fig. 7). For the oceanic areas, 41 %  
20 ( $0.086\pm 0.085$ ) of the total AOD<sub>550</sub> is due to anthropogenic aerosols, 34 % ( $0.076\pm 0.185$ ) due  
21 to dust and 25 % ( $0.054\pm 0.018$ ) due to marine aerosols. The results based on observations  
22 from MODIS Aqua are similar to the MODIS Terra ones and in accord with previous studies.

23 Over land, the  $\tau_a$  maxima are detected over local particle pollution sources (cities, industrial  
24 zones, etc.). Over the sea,  $\tau_a$  is higher along the coasts being significantly lower at greater  
25 distance. Very high  $\tau_d$  values appear over land regions in North Africa, Middle East, Anatolia  
26 and oceanic areas across the Eastern Mediterranean, especially for latitudes below 35°N. Over  
27 the sea, dust loading exhibits two maxima, one at the coastal zone of Libya and one across the  
28 coastal zone of the Middle East.  $\tau_d$  decreases with distance from the large dust sources in the  
29 South and South-East. Generally, dust reaches heights up to  $\sim 4-5$  km in the area, the largest  
30 fraction of dust mass being confined within the first 2-3 km of the troposphere. The spatial  
31 variability of  $\tau_n$  and  $\tau_m$  is very low compared to  $\tau_a$  and  $\tau_d$ , following the total AOD<sub>550</sub> patterns  
32 and the near surface wind speed patterns, respectively.

1 Over land, the relative contribution of anthropogenic aerosols, dust and fine mode natural  
2 aerosols to the total AOD<sub>550</sub> exhibits a low seasonal variability, while over the sea the relative  
3 contribution of anthropogenic aerosols, dust and marine aerosols shows a significant seasonal  
4 variability.

5  $\tau_a$  nearly doubles during the warm period of the year (spring-summer), August and April  
6 being the months with the highest  $\tau_a$  levels. The summer peak is mostly the result of low  
7 precipitation levels, enhancement of the photochemical production of secondary organic  
8 aerosols and transport of pollution aerosols from Central Europe and biomass burning  
9 aerosols from South-Western Russia and Eastern Europe during the biomass burning season  
10 in July-August. The spring maximum in April is mostly the result of transport of biomass  
11 burning aerosols from Russia in line with previous studies. Dust exhibits a strong peak in  
12 spring (April), especially over the southern regions. April is the month with the highest  $\tau_a$   
13 levels as a result of the high cyclonic activity over North Africa. The seasonal variability of  
14 dust is much stronger and the spring maxima much more prominent over the sea as dust is  
15 only occasionally transported there during episodic events, while over land, local sources  
16 contribute to the dust burden, especially in summer due to the soil dryness. The seasonal  
17 variability of fine mode natural aerosols is very low, the highest values appearing in spring  
18 and summer. Marine aerosols also present a very low seasonal variability, the highest values  
19 appearing in winter due to the high near surface wind speeds.

20 Overall, it is suggested that the AOD<sub>550</sub>,  $\tau_a$ ,  $\tau_d$ ,  $\tau_n$  and  $\tau_m$  high resolution gridded dataset which  
21 was compiled in this work could be used in a number of future atmospheric and biological  
22 studies focusing on the region of the Eastern Mediterranean (e.g. satellite and ground-based  
23 studies on aerosol-cloud-radiation interactions, experimental and field campaign studies on  
24 aerosols and clouds and research on the impact of aerosols on human health and nature). It is  
25 also acknowledged that a future update of the results presented here using more recent  
26 releases of MODIS aerosol data (e.g. Collection 6) and aerosol reanalysis datasets (e.g.  
27 NASA's Modern-Era Retrospective Analysis For Research And Applications Aerosol Re-  
28 analysis) would be a useful contribution.

29

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23

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1 **Table 1.** Full name, abbreviation, geolocation, host country and type of the 13 AERONET  
 2 Cimel sunphotometer sites used for the validation of MODIS Terra and Aqua Collection 051  
 3 observations. The common measurement period of MODIS and AERONET data and the  
 4 corresponding overpass time of MODIS Terra and Aqua (*Italics*) over each station are also  
 5 given.

AERONET Station	Lat (°N)	Lon (°E)	Period of study	Country	Type	TERRA overpass	<i>AQUA overpass</i>
ATHENS-NOA (ATH)	37.988	23.775	05/2008-10/2012	Greece	Urban (coastal)	9:23±22min UT	<i>11:32±22min UT</i>
Bucharest Inoe (BUC)	44.348	26.030	07/2007-09/2012	Romania	Sub-urban (coastal)	9:17±24min UT	<i>11:15±20min UT</i>
CUT-TEPAK (CUT)	34.675	33.043	04/2010-12/2012	Cyprus	Urban (coastal)	8:43±25min UT	<i>10:55±25min UT</i>
Eforie (EFO)	44.075	28.632	09/2009/12/2012	Romania	Rural (coastal)	9:09±21min UT	<i>11:04±21min UT</i>
FORTH Crete (FOR)	35.333	25.282	01/2003-08/2011	Greece	Rural (coastal)	9:12±24min UT	<i>11:25±25min UT</i>
IMS-METU-ERDEMLI (IMS)	36.565	34.255	01/2004-01/2012	Turkey	Rural (coastal)	8:39±23min UT	<i>10:48±22min UT</i>
Lecce University (LEC)	40.335	18.111	03/2003-12/2012	Italy	Sub-urban (coastal)	9:44±25min UT	<i>11:49±25min UT</i>
Nes ziona (NES)	31.922	34.789	02/2000-12/2012	Israel	Sub-urban (coastal)	8:38±24min UT	<i>10:44±25min UT</i>
SEDE BOKER (SED)	30.855	34.782	01/2000-04/2012	Israel	Rural (semi-arid)	8:30±27min UT	<i>10:50±25min UT</i>
Sevastopol (SEV)	44.616	33.517	05/2006-12/2012	Ukr.-Crimea	Urban (coastal)	8:51±21min UT	<i>10:40±21min UT</i>
Thessaloniki (THE)	40.630	22.960	09/2005-12/2012	Greece	Urban (coastal)	9:28±25min UT	<i>11:32±22min UT</i>
TUBITAK UZAY Ankara (TUB)	39.891	32.778	12/2009-04/2012	Turkey	Urban (continental)	8:48±26min UT	<i>10:56±24min UT</i>
Xanthi (XAN)	41.147	24.919	01/2008-10/2010	Greece	Rural (coastal)	9:18±25min UT	<i>11:24±21min UT</i>

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1 **Table 2.** Results of the comparison of spatially (using a spatial window around each station)  
2 and temporally ( $\pm 30$  min from the MODIS overpass time) collocated MODIS Terra and Aqua  
3 (Italics) Collection 051 level-2 and AERONET sunphotometric (quadratically interpolated)  
4 AOD<sub>550</sub> observations for the Eastern Mediterranean stations. The algorithms used for the  
5 production of the validated MODIS data (DT and DB), the spatial window used for the spatial  
6 collocation (25 x 25 km<sup>2</sup> or 50 x 50 km<sup>2</sup> window around each station) with the AERONET  
7 data, the average MODIS and AERONET AOD<sub>550</sub> and the corresponding  $\pm 1\sigma$  values, the  
8 mean difference between them, the normalized mean bias (NMB) and the corresponding root  
9 mean square (RMS) error, the percentage of the collocation points that fall within the  
10 expected error (EE) envelope and the pre-launch expected error (pLEE) envelope (Expected  
11 Uncertainty - EU envelope for DB data), the correlation coefficient R, the slope (a) and the  
12 intercept of the regression line (b) and the number of the collocation points are given in the  
13 table. L10 denotes the use of a collocation window of 50° x 50° as in Levy et al. (2010) while  
14 HQ denotes the use of high quality data only.

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Alg.	Window	MODIS TERRA <i>MODIS AQUA</i>	AERONET	Mean Diff.	NMB %	RMS err.	in EE %	in pl EE %	R	a	b	Obs
DT	25 km	0.223±0.163	0.200±0.123	0.023±0.106	11.59	0.11	63.28	67.78	0.76	1.007	0.022	6697
<i>DT</i>	<i>25 km</i>	<i>0.247±0.173</i>	<i>0.197±0.121</i>	<i>0.050±0.109</i>	<i>25.18</i>	<i>0.12</i>	<i>57.14</i>	<i>61.87</i>	<i>0.78</i>	<i>1.113</i>	<i>0.027</i>	<i>6283</i>
DT	50 km (L10)	0.204±0.152	0.194±0.124	0.010±0.085	5.10	0.09	70.17	74.64	0.83	1.016	0.007	6054
<i>DT</i>	<i>50 km (L10)</i>	<i>0.224±0.155</i>	<i>0.194±0.125</i>	<i>0.030±0.088</i>	<i>15.34</i>	<i>0.09</i>	<i>66.76</i>	<i>70.45</i>	<i>0.82</i>	<i>1.018</i>	<i>0.026</i>	<i>5557</i>
DB	25 km	0.226±0.177	0.186±0.128	0.040±0.162	21.38	0.17	-	51.90	0.47	0.657	0.104	2580
<i>DB</i>	<i>25 km</i>	<i>0.242±0.217</i>	<i>0.182±0.118</i>	<i>0.06±0.196</i>	<i>33.03</i>	<i>0.20</i>	-	<i>55.30</i>	<i>0.44</i>	<i>0.815</i>	<i>0.094</i>	<i>5345</i>
DB <sub>HQ</sub>	25 km	0.229±0.158	0.186±0.132	0.043±0.141	22.82	0.15	-	52.41	0.54	0.651	0.108	498
<i>DB<sub>HQ</sub></i>	<i>25 km</i>	<i>0.260±0.220</i>	<i>0.186±0.138</i>	<i>0.074±0.204</i>	<i>39.84</i>	<i>0.22</i>	-	<i>52.34</i>	<i>0.42</i>	<i>0.670</i>	<i>0.136</i>	<i>896</i>

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1 **Table 3.** AOD<sub>550</sub> levels, the corresponding  $\pm 1\sigma$  values and the number of gridded values used  
 2 for the calculations over the Eastern Mediterranean (EMT), over the land covered part (EML),  
 3 over the oceanic part and over the 9 sub-regions of the Eastern Mediterranean appearing in  
 4 Fig. 1 based on the MODIS Terra and Aqua (*Italics*) observations.

Region	MODIS TERRA AOD <sub>550</sub>	Num. of values	<i>MODIS AQUA</i> <i>AOD<sub>550</sub></i>	Num. of values
EMT	0.215±0.187	61496654	<i>0.217±0.199</i>	<i>49522934</i>
EML	0.219±0.165	25923766	<i>0.239±0.189</i>	<i>21008713</i>
EMO	0.213±0.201	35572888	<i>0.202±0.205</i>	<i>28514221</i>
NBL	0.183±0.163	5563495	<i>0.187±0.162</i>	<i>3853688</i>
SBL	0.197±0.152	7345829	<i>0.207±0.152</i>	<i>5272449</i>
ANL	0.223±0.146	7948817	<i>0.228±0.148</i>	<i>5539261</i>
NAL	0.282±0.192	5065625	<i>0.306±0.238</i>	<i>6343315</i>
BSO	0.198±0.150	6433951	<i>0.183±0.134</i>	<i>5262438</i>
NWO	0.209±0.162	11645069	<i>0.197±0.154</i>	<i>9231630</i>
SWO	0.226±0.266	6202893	<i>0.223±0.310</i>	<i>4925665</i>
NEO	0.214±0.196	4807910	<i>0.199±0.166</i>	<i>3896554</i>
SEO	0.221±0.236	6483065	<i>0.210±0.239</i>	<i>5197934</i>

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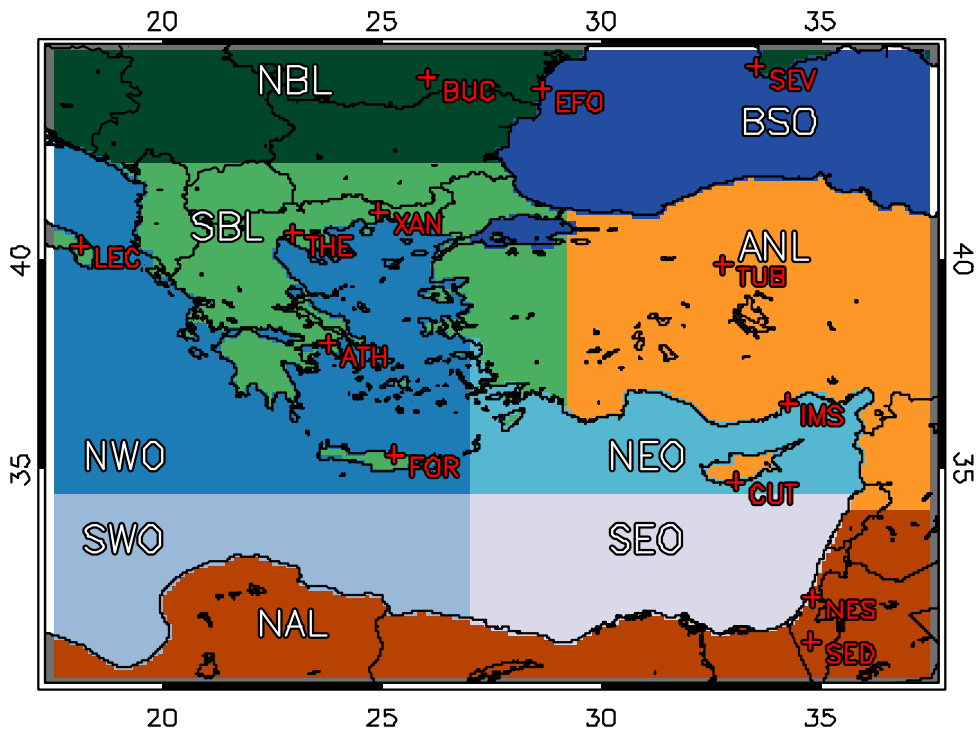
1 **Table 4.** Relative contribution of anthropogenic aerosols, dust, fine mode natural and marine  
2 aerosols to the total AOD<sub>550</sub> (bold) and the corresponding  $\tau_a$ ,  $\tau_d$ ,  $\tau_n$ ,  $\tau_m$  levels with their  $\pm 1\sigma$   
3 values (in parentheses) over the Eastern Mediterranean (EMT), over the land covered part  
4 (EML), over the oceanic part and over the 9 sub-regions of the Eastern Mediterranean  
5 appearing in Fig. 1 based on the MODIS Terra and Aqua (Italics) observations. The sum of  
6 the aerosol type AODs per region does not necessarily correspond to the total AOD<sub>550</sub> values  
7 appearing in Table 3 as these results were for the total of the days with aerosol retrievals even  
8 for days when our aerosol type separation algorithm was not applicable.

Region	Satellite	Contribution to MODIS TERRA/AQUA AOD <sub>550</sub>			
		Anthropogenic	Dust	Fine mode natural	Marine
EML	TERRA	<b>52 %</b> (0.112±0.087)	<b>32 %</b> (0.074±0.080)	<b>16 %</b> (0.034±0.026)	-
	AQUA	<b>50 %</b> (0.117±0.093)	<b>35 %</b> (0.090±0.102)	<b>15 %</b> (0.035±0.028)	-
EMO	TERRA	<b>41 %</b> (0.086±0.085)	<b>34 %</b> (0.076±0.185)	-	<b>25 %</b> (0.054±0.018)
	AQUA	<b>40 %</b> (0.079±0.080)	<b>33 %</b> (0.070±0.181)	-	<b>27 %</b> (0.054±0.018)
NBL	TERRA	<b>59 %</b> (0.108±0.101)	<b>23 %</b> (0.042±0.046)	<b>18 %</b> (0.032±0.030)	-
	AQUA	<b>59 %</b> (0.110±0.100)	<b>24 %</b> (0.045±0.047)	<b>17 %</b> (0.033±0.030)	-
SBL	TERRA	<b>55 %</b> (0.109±0.088)	<b>28 %</b> (0.056±0.058)	<b>17 %</b> (0.033±0.026)	-
	AQUA	<b>55 %</b> (0.113±0.088)	<b>29 %</b> (0.060±0.060)	<b>16 %</b> (0.034±0.026)	-
ANL	TERRA	<b>51 %</b> (0.113±0.075)	<b>34 %</b> (0.076±0.068)	<b>15 %</b> (0.034±0.023)	-
	AQUA	<b>50 %</b> (0.114±0.075)	<b>35 %</b> (0.079±0.070)	<b>15 %</b> (0.034±0.023)	-
NAL	TERRA	<b>50 %</b> (0.113±0.083)	<b>35 %</b> (0.083±0.085)	<b>15 %</b> (0.034±0.025)	-
	AQUA	<b>48 %</b> (0.118±0.091)	<b>38 %</b> (0.099±0.108)	<b>14 %</b> (0.035±0.027)	-
BSO	TERRA	<b>53 %</b> (0.108±0.103)	<b>22 %</b> (0.044±0.101)	-	<b>25 %</b> (0.051±0.016)
	AQUA	<b>51 %</b> (0.094±0.087)	<b>22 %</b> (0.042±0.085)	-	<b>27 %</b> (0.051±0.016)
NWO	TERRA	<b>41 %</b> (0.087±0.090)	<b>33 %</b> (0.071±0.142)	-	<b>26 %</b> (0.055±0.020)
	AQUA	<b>40 %</b> (0.079±0.083)	<b>32 %</b> (0.066±0.127)	-	<b>28 %</b> (0.055±0.020)
SWO	TERRA	<b>32 %</b> (0.071±0.070)	<b>42 %</b> (0.097±0.257)	-	<b>26 %</b> (0.058±0.018)
	AQUA	<b>32 %</b> (0.093±0.288)	<b>41 %</b> (0.072±0.080)	-	<b>27 %</b> (0.059±0.018)
NEO	TERRA	<b>48 %</b> (0.098±0.094)	<b>28 %</b> (0.061±0.144)	-	<b>24 %</b> (0.050±0.016)
	AQUA	<b>46 %</b> (0.086±0.082)	<b>28 %</b> (0.057±0.115)	-	<b>26 %</b> (0.050±0.016)
SEO	TERRA	<b>36 %</b> (0.079±0.070)	<b>39 %</b> (0.087±0.224)	-	<b>25 %</b> (0.055±0.016)
	AQUA	<b>36 %</b> (0.075±0.071)	<b>38 %</b> (0.080±0.217)	-	<b>26 %</b> (0.055±0.016)

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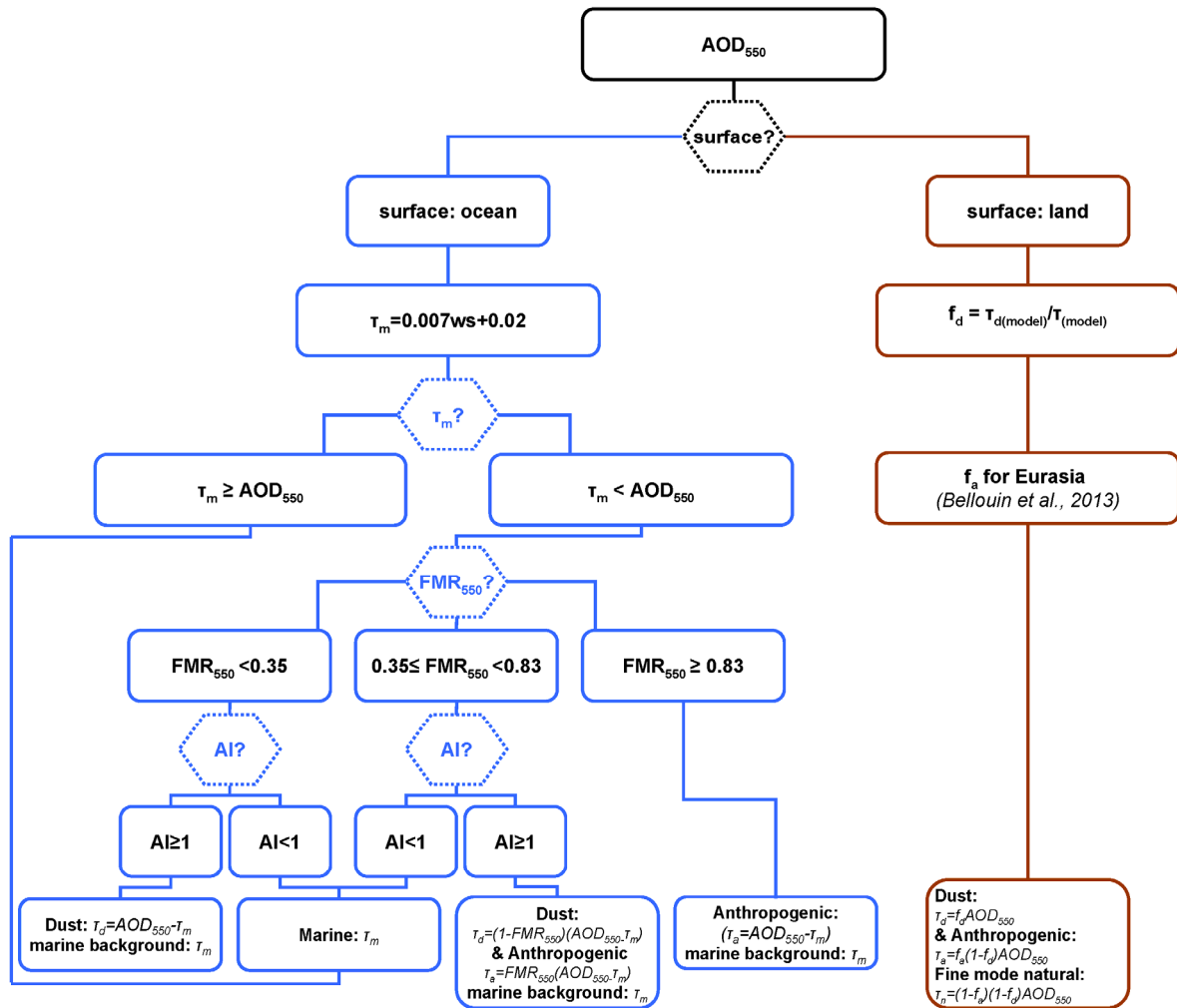


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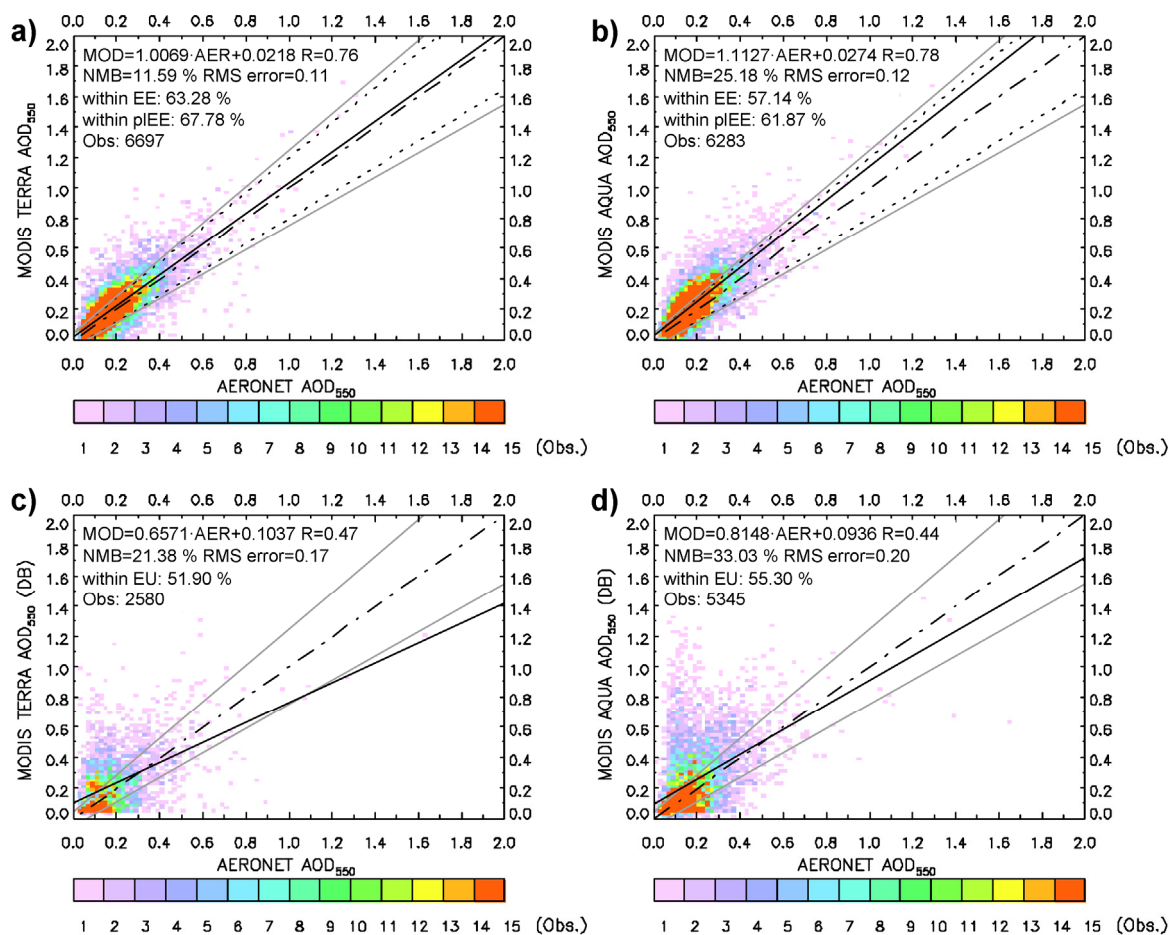
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3 **Figure 1.** Eastern Mediterranean map with the 9 sub-regions selected for the generalization of  
4 our results and the location of the AERONET stations used for the validation of MODIS  
5 satellite data. The 9 sub-regions are: NBL (Northern Balkans Land), SBL (Southern Balkans  
6 Land), ANL (Anatolia Land), NAL (Northern Africa Land), BSO (Black Sea Oceanic), NWO  
7 (North-Western Oceanic), SWO (South-Western Oceanic), NEO (North-Eastern Oceanic) and  
8 SEO (South-Eastern Oceanic). The full names and the geolocation of the 13 AERONET  
9 stations appearing in the map are available in Table 1.



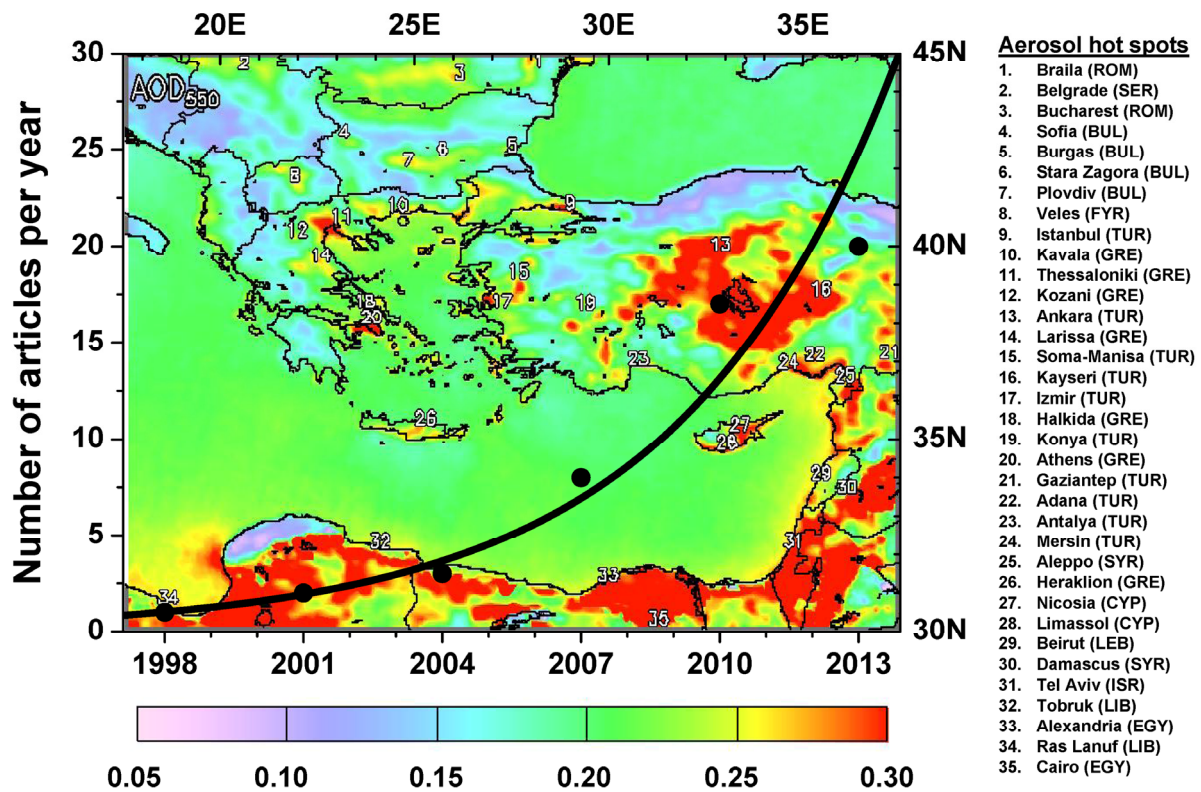
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**Figure 2.** Flowchart with the methodology followed for the calculation of the anthropogenic aerosol, dust and marine aerosol optical depths ( $\tau_a$ ,  $\tau_d$  and  $\tau_m$ ) over the sea (blue color) and the anthropogenic aerosol, dust and fine mode natural aerosol optical depths ( $\tau_a$ ,  $\tau_d$  and  $\tau_n$ ) over land (brown color).

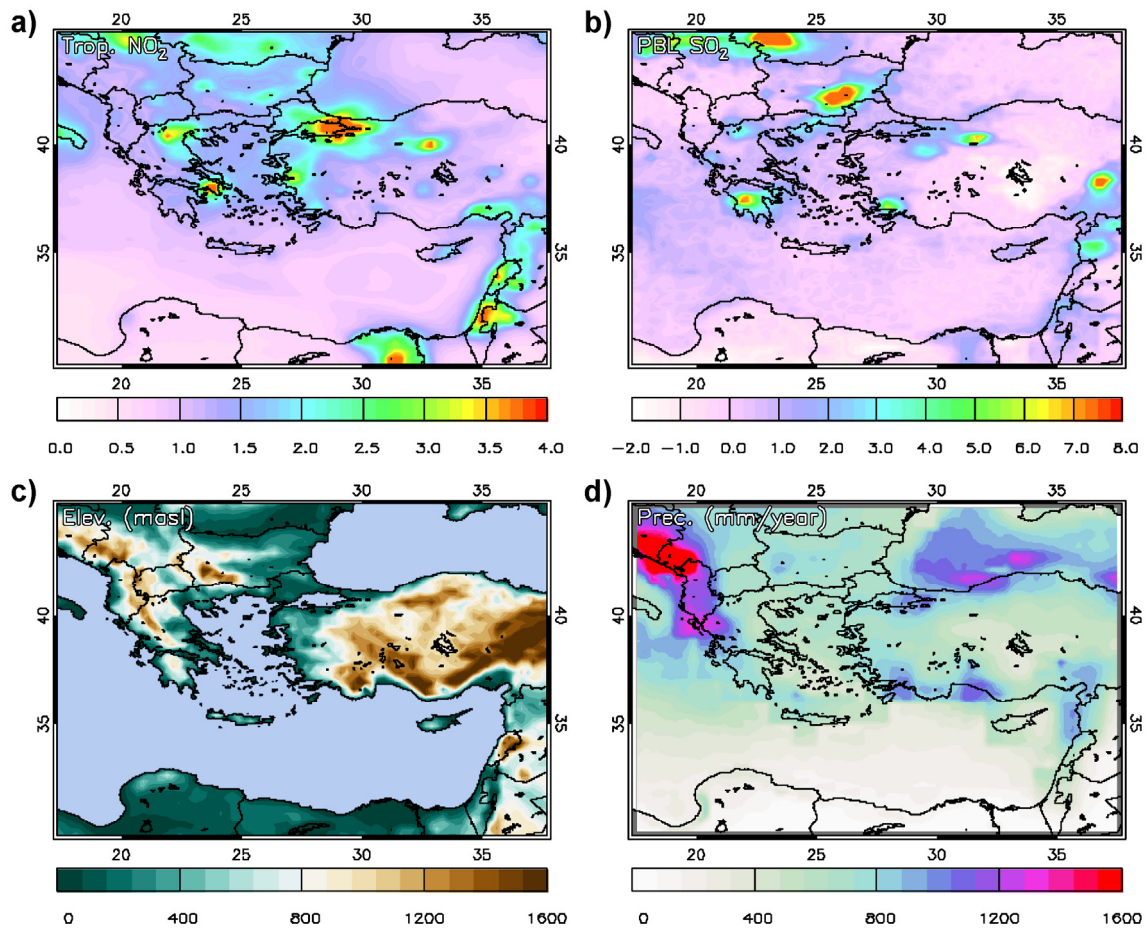


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3 **Figure 3.** Comparison of spatially (using a 25 x 25 km<sup>2</sup> window around each station) and  
4 temporally ( $\pm 30$  min from the MODIS overpass time) collocated MODIS Collection 051  
5 level-2 and AERONET sunphotometric (quadratically interpolated) AOD<sub>550</sub> observations for  
6 the Eastern Mediterranean stations: (a) for MODIS Terra DT data, (b) for MODIS Aqua DT  
7 data, (c) for MODIS Terra DB data and (d) for MODIS Aqua DB data. The color scale  
8 corresponds to the number of MODIS-AERONET collocation points that fall within 0.02 x  
9 0.02 grid boxes. The solid line is the regression line of the MODIS-AERONET observations,  
10 the dashed-dotted line is the 1:1 line, the dotted lines represent the expected error (EE)  
11 envelope and the grey lines the pre-launch expected error (pLEE) envelope (Expected  
12 Uncertainty - EU envelope for DB data). The slope and the intercept of the regression line, the  
13 correlation coefficient R, the normalized mean bias (NMB), the root mean square (RMS)  
14 error, the percentage of the collocation points that fall within the EE and pLEE and the number  
15 of all the collocation points are given on the plots.

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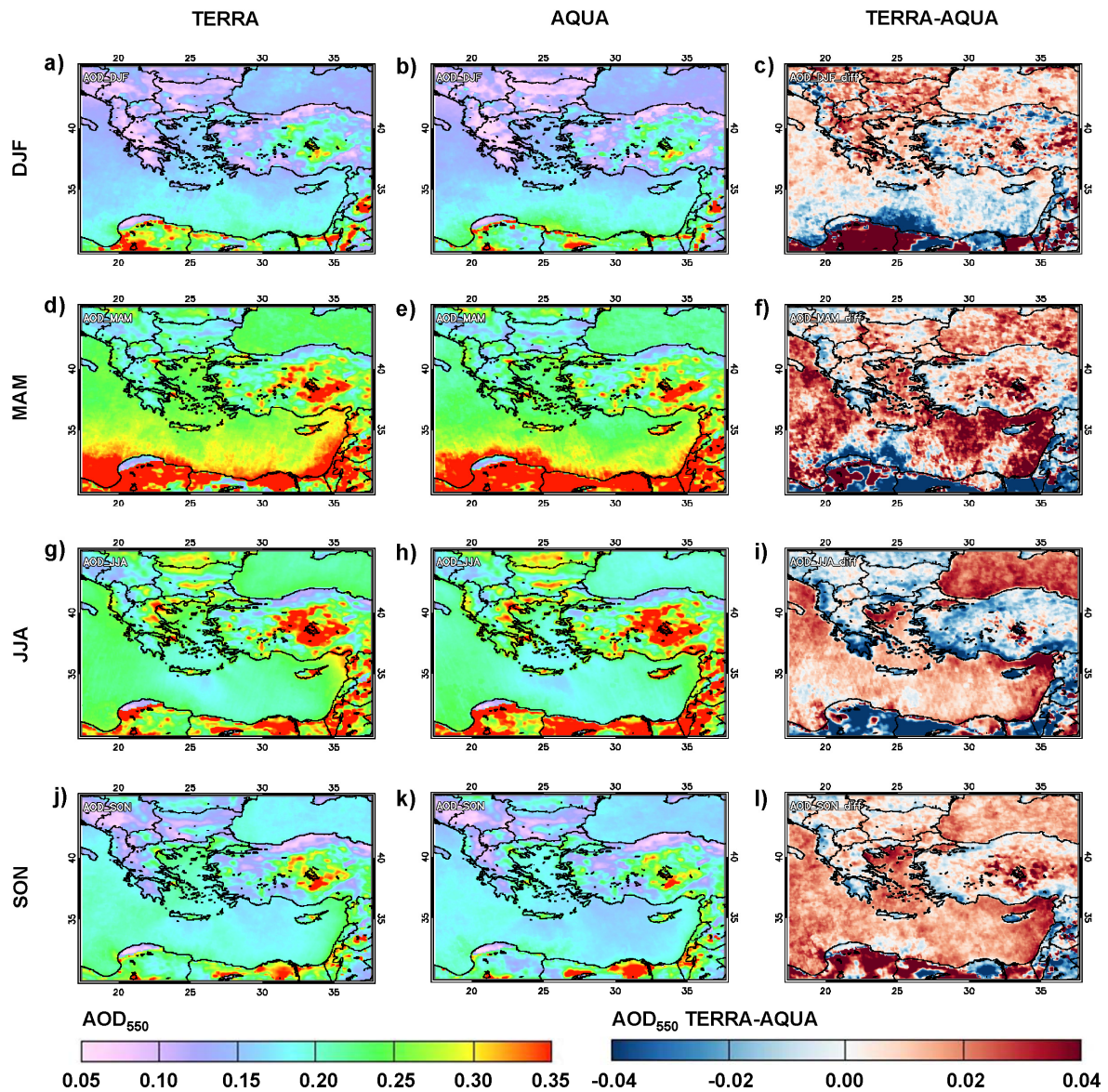


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3 **Figure 4.** AOD<sub>550</sub> patterns over Eastern Mediterranean as seen by MODIS Terra during the  
4 period 3/2000-12/2012 (3/2000-12/2007 for regions of North Africa covered by DB data  
5 only). The colorscale corresponds to the AOD<sub>550</sub> levels while the top x-axis and the right y-  
6 axis correspond to the longitude (°E) and latitude (°N), respectively. The position of 35  
7 aerosol hot spots is marked on the map (numbers from 1 to 35) while the names of the places  
8 and the countries where the hot spots are located appear on the right of the map. In the same  
9 figure the exponential growth of the number of satellite-based articles focusing on aerosols  
10 over the greater Eastern Mediterranean from 1997 to 2014 is shown (black line). The black  
11 dots represent the number of articles published within three year intervals. The bottom x-axis  
12 and the left y-axis correspond to the years and the number of published articles, respectively.  
13 The exponential growth corresponds to a near doubling of the publication rate every 3 years.



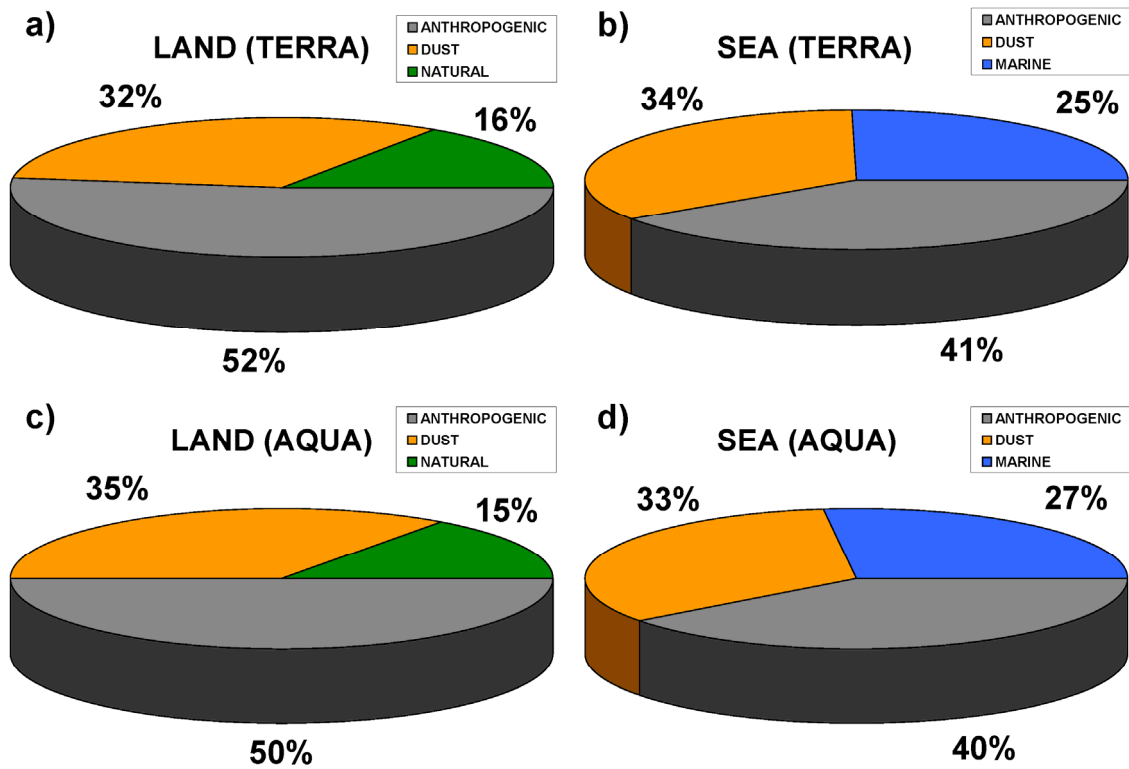
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3 **Figure 5.** (a) Tropospheric NO<sub>2</sub> levels and (b) Planetary boundary layer SO<sub>2</sub> levels (in 10<sup>15</sup>  
 4 molecules/cm<sup>2</sup>) over the Eastern Mediterranean as seen from OMI/AURA (2005-2012), (c)  
 5 Topography (GTOPO elevation data in meters above sea level) and (d) Annual precipitation  
 6 levels (in mm/year) from 3B43 TRMM and Other Sources Monthly Rainfall Product (2000-  
 7 2012).



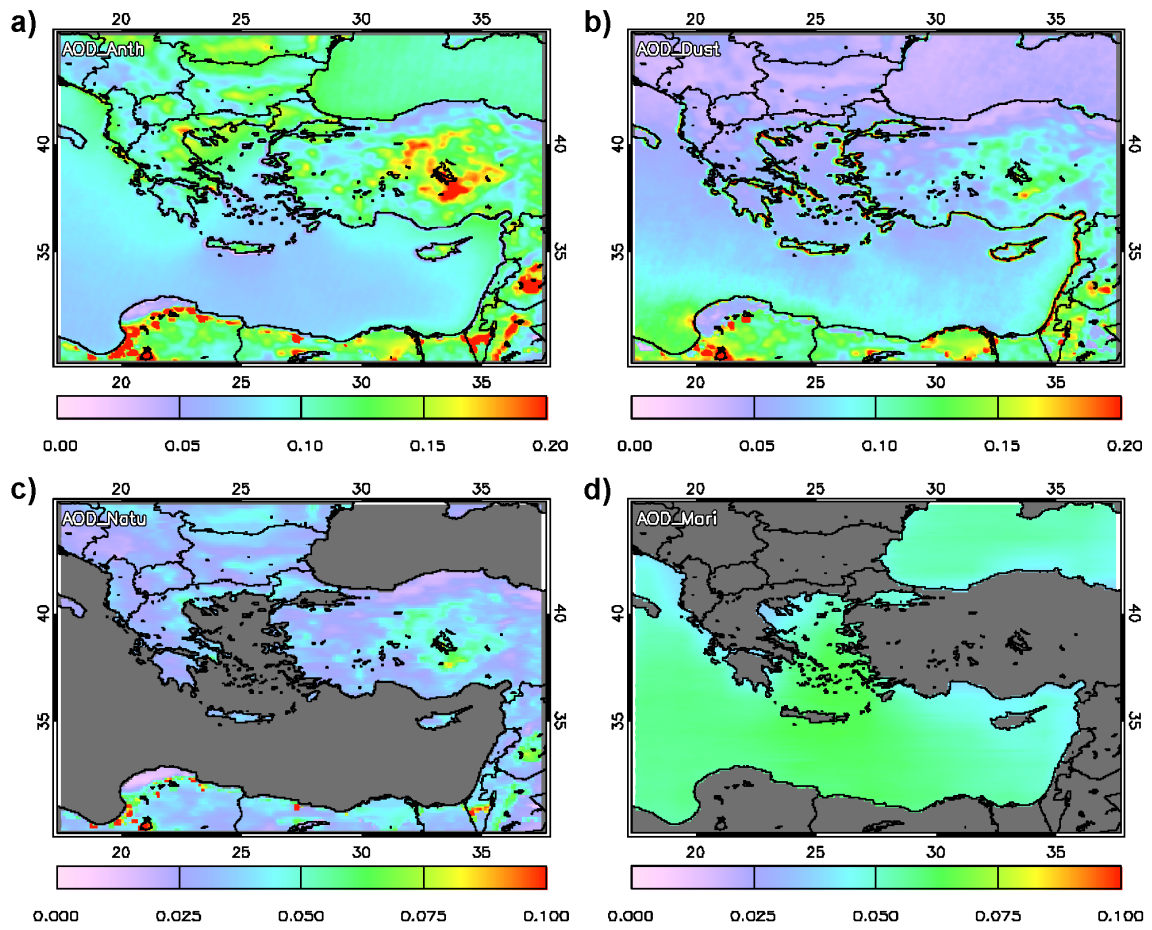
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**Figure 6.** Seasonal AOD<sub>550</sub> patterns over the Eastern Mediterranean as seen by MODIS Terra (left column) during the period 3/2000-12/2012 (3/2000-12/2007 for regions of North Africa covered by DB data only) and MODIS Aqua (middle column) during the period 7/2002-12/2012. The differences between MODIS Terra and Aqua AOD<sub>550</sub> on a seasonal basis appear on the right column.



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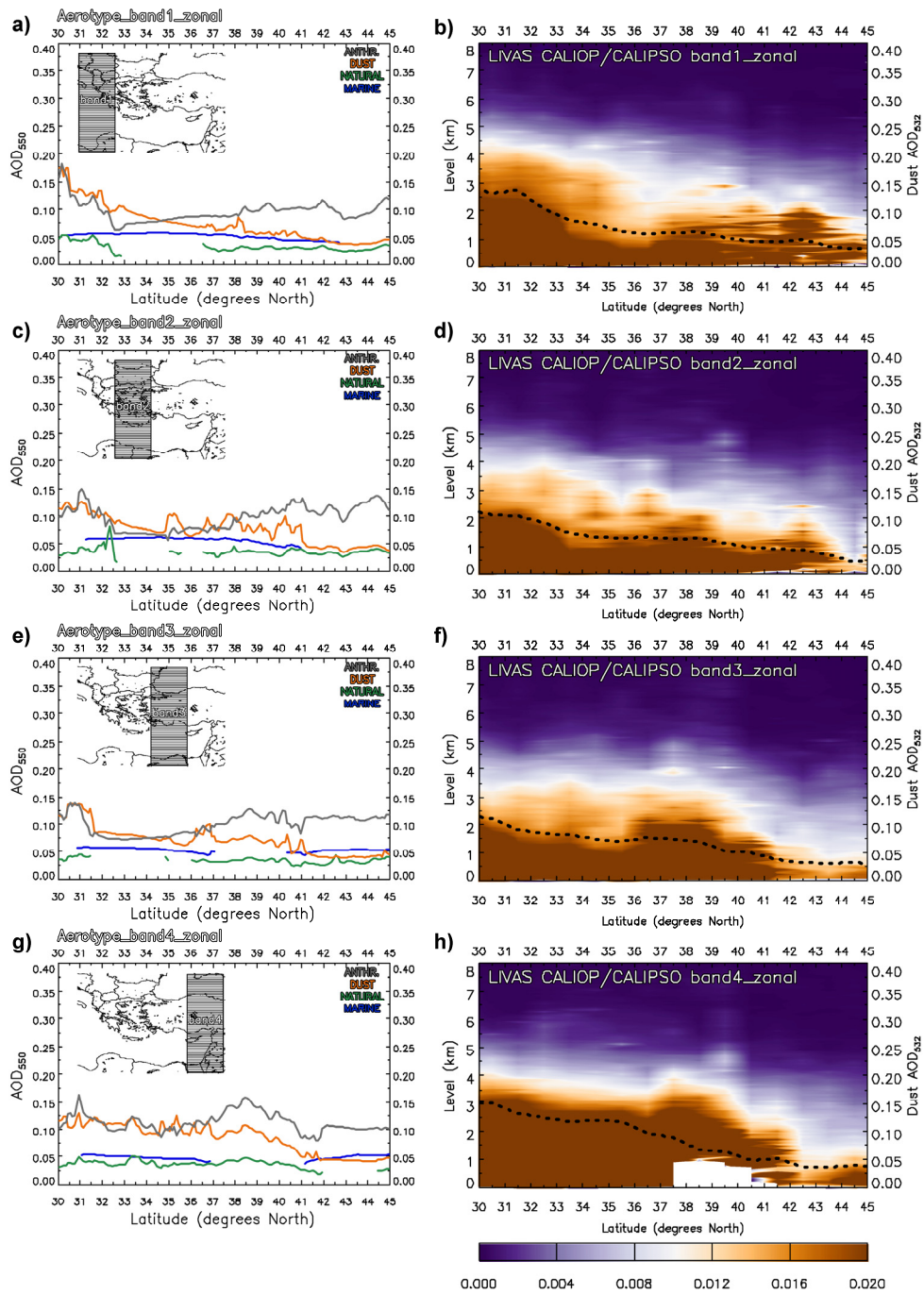
**Figure 7.** Relative contribution of anthropogenic aerosols, dust and fine mode natural aerosols to the total AOD<sub>550</sub> over the land covered part of the Eastern Mediterranean based on MODIS Terra (a) and MODIS Aqua (c) observations and relative contribution of anthropogenic aerosols, dust and marine aerosols to the total AOD<sub>550</sub> over the oceanic part of the Eastern Mediterranean based on MODIS Terra (b) and MODIS Aqua (d) observations.



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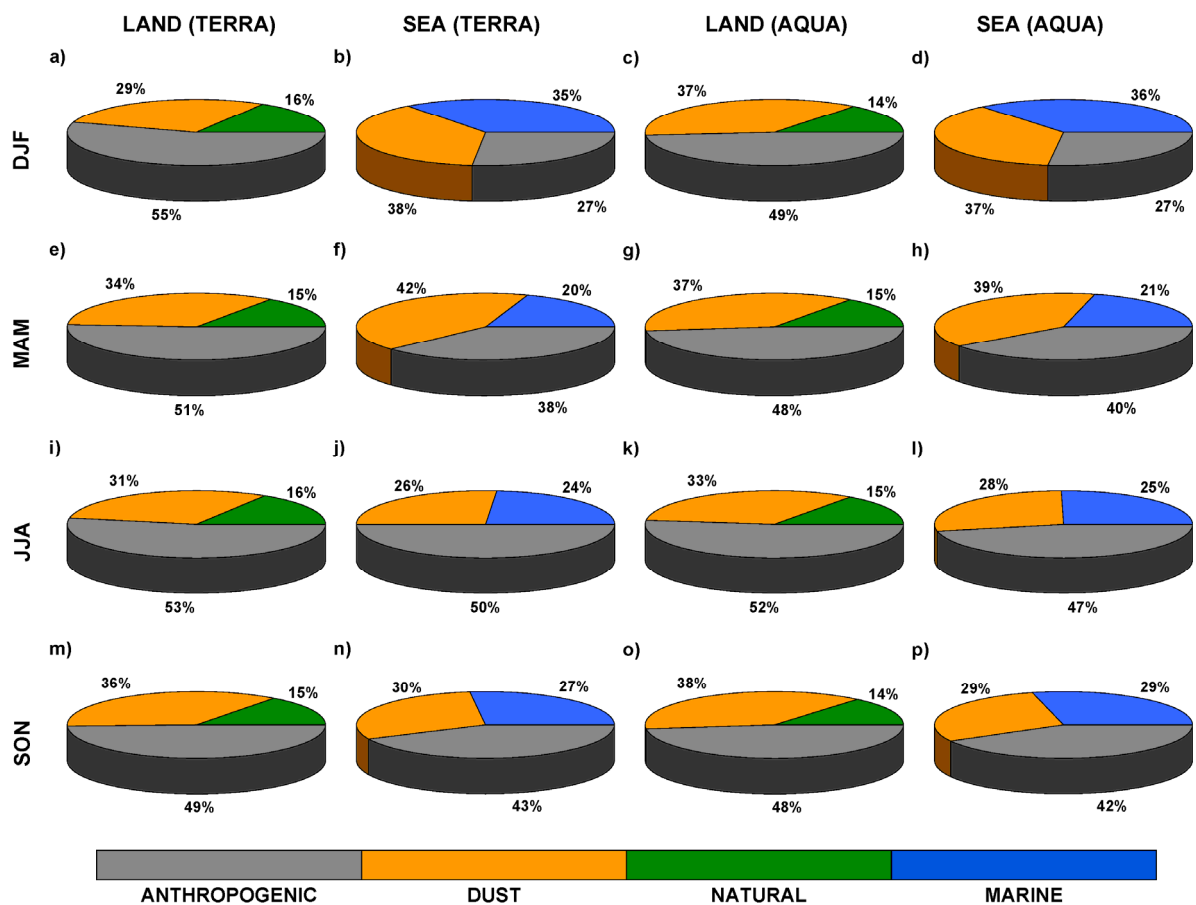
3 **Figure 8.** (a) Anthropogenic aerosol ( $\tau_a$ ), (b) dust ( $\tau_d$ ), (c) fine mode natural aerosol ( $\tau_n$ ) and  
 4 (d) marine aerosol ( $\tau_m$ ) patterns over the Eastern Mediterranean based on MODIS Terra  
 5 observations during the period 3/2000-12/2012 (3/2000-12/2007 for regions of North Africa  
 6 covered by DB data only).





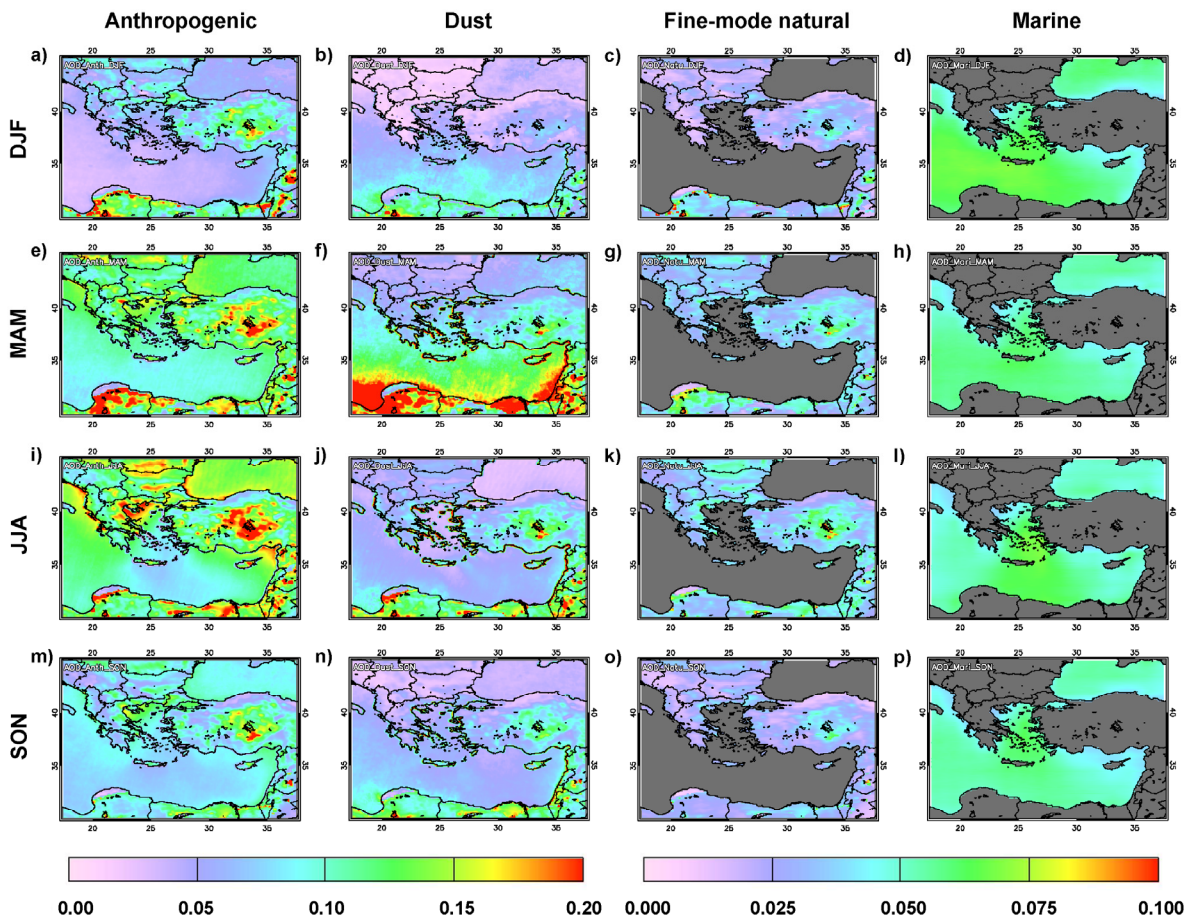
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**Figure 9.** Left column: Latitudinal variability of anthropogenic aerosols ( $\tau_a$ ), dust ( $\tau_d$ ), fine mode natural aerosols ( $\tau_n$ ) and marine aerosols ( $\tau_m$ ) for four 5-degree longitudinal bands (see embedded maps) covering the Eastern Mediterranean based on MODIS Terra observations. Right column: Latitudinal variability of dust extinction coefficients at 532 nm in  $\text{km}^{-1}$  (color scale corresponds to the extinction coefficients and left y-axis to the atmospheric levels) and dust aerosol optical depth at 532 nm (dotted line corresponding to the right y-axis) for the same four bands from LIVAS CALIOP/CALIPSO observations.



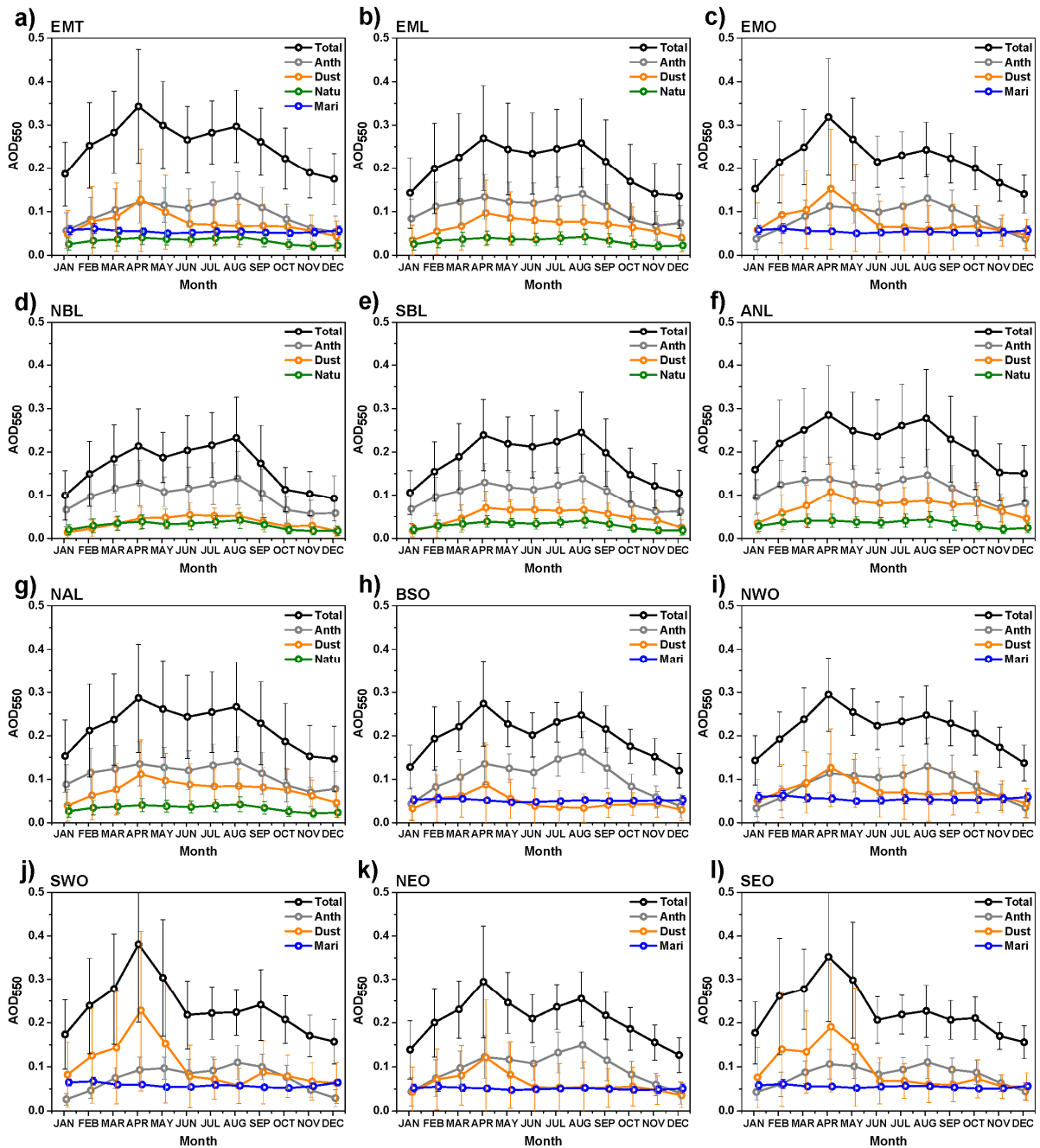
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**Figure 10.** Seasonal relative contribution of anthropogenic aerosols, dust and fine mode natural aerosols to the total AOD<sub>550</sub> over the land covered part of the Eastern Mediterranean based on MODIS Terra (a, e, i, m) and MODIS Aqua (c, g, k, o) observations and seasonal relative contribution of anthropogenic aerosols, dust and marine aerosols to the total AOD<sub>550</sub> over the oceanic part of the Eastern Mediterranean based on MODIS Terra (b, f, j, n) and MODIS Aqua (d, h, l, p) observations.



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3 **Figure 11.** Seasonal (a, e, i, m) anthropogenic aerosol ( $\tau_a$ ), (b, f, j, n) dust ( $\tau_d$ ), (c, g, k, o) fine  
4 mode natural aerosol ( $\tau_n$ ) and (d, h, i, p) marine aerosol ( $\tau_m$ ) patterns over the Eastern  
5 Mediterranean based on MODIS Terra observations during the period 3/2000-12/2012  
6 (3/2000-12/2007 for regions of North Africa covered by DB data only).



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 3 **Figure 12.** Seasonal variability of anthropogenic aerosols ( $\tau_a$ ), dust ( $\tau_d$ ), fine mode natural  
 4 aerosols ( $\tau_n$ ) and marine aerosols ( $\tau_m$ ) over the Eastern Mediterranean (EMT), over the land  
 5 covered part (EML), over the oceanic part (EMO) and over the 9 sub-regions of the Eastern  
 6 Mediterranean appearing in Fig. 1 based on MODIS Terra observations. The error bars  
 7 represent the  $\pm 1\sigma$  values calculated from monthly gridded data.

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