Response to Reviewers

The authors would like to thank the Anonymous Reviewers for their comments. Below, please find attached our response to each one of the reviewers' comments:

Reviewer #1

1. Line 1, Page 2: AERONET provide the full name.

We have added the AERONET full name in the revised version of the manuscript.


This has been corrected in the revised version of the manuscript.

3. Line 7, Page 3: Please add some references for the sentence "The Mediterranean is also recognized as a crossroads between three continents where aerosols of various types accumulate”.

The classic paper from Lelieveld et al. (2002) that inserted the term "crossroads" for the Mediterranean Basin has been added.

4. Line 1, Page 4: The references format should be unified, for example: “Gkikas et al., 2013; 2014;” should be “Gkikas et al., 2013, 2014”., etc. Please revise all similar reference formats through the full text.

We have corrected this wherever it appeared in the manuscript.

5. Line 7, Page 7: “AE440-870”should be “AE440-870”.

This has been corrected in the revised version of the manuscript.

6. Line 11, Page 8: “resolution of 1° x 1.25” : Please add the note which is latitude? Which is longitude?

We write it like this "1o (latitude) x 1.25o (longitude) " in the revised version of the manuscript.

7. Line 32, Page 8: What is the resolution of wind speed data at 10 m above surface from ERA-Interim reanalysis used in your paper?

We thank the reviewer for giving use the opportunity to clarify this. We added the following in the revised version of the manuscript "...the data can be acquired at various resolutions (in this work 1ox1o) through ECMWF’s website..."

8. Why have the years after 2012 been excluded?

We refer in the revised manuscript that the years after 2012 have been excluded as the MACC reanalysis data are only available for the period 2003-2012.
9. AERONET AOD$_{550}$ is how calculated?

We had already written about this in the paper "... Here, we use quadratic fits on a log-log scale to interpolate the AERONET data (AODs at 440, 500, 675 and 870 nm) to the MODIS band-effective wavelength of 550 nm (Eck et al., 1999; Levy et al., 2010)..."

However, we also added the following phrase in order to make it more clear "...So, we can directly compare the MODIS AOD$_{550}$ retrievals against AERONET observations."

10. Line 12-13, Page 21: Please add some references for the sentence "Over the ocean, a profound maximum is observed in spring which is due to the well documented transport of significant amounts of dust from the Sahara Desert extending across the North African coast and the ocean."

As this spring transport is found in numerous papers about the area we added the following in the sentence "...(see Barnaba and Gobbi., 2004 and the list of references given in the introduction)...

11. Line 3, Page 25: “from 49 % in SON to 50 %” should be “from 49 % in SON to 55 %”.

This has been corrected in the revised version of the manuscript.

12. Figure 6d: AOD$_{550}$ at NAL region in spring is bigger than it in summer. Why?

This is mostly due to the significant amount of dust produced and transported in this region during spring. This can also be seen in Fig. 11 where we see the significantly higher amounts of dust over the region in spring compared to summer.

13. The conclusion is a very long summary, whether can be shorten appropriately?

We agree with the reviewer that the conclusions section is pretty long. So we decided to rename this section to Summary and conclusions. As this is a long paper and it is very common among scientists to focus on the abstract and the conclusions we prefer to keep the Summary and conclusions section as informative as possible (numbers, conclusions, methodology, future use of the dataset, etc.).

14. Whether to consider joining the back trajectory model to track the source of dust?

We thank the reviewer for his/her comment. We have already plans to extend the methodology as discussed in the Conclusions using more recent datasets that will allow the extension of the timeseries and we also are working on possible improvements of the methodology. Indeed the use of trajectory models could be useful especially over the sea as over land the method relies on data from models and the trajectories are not expected to return significantly different results. One of our future goals is to separate cases that dust is transported from Sahara from case that dust comes from Middle East and trajectory analysis will play a critical role.
Reviewer #2

1. insert ‘the’ before ‘Mediterranean’ throughout the whole MS.

We have addressed this in the revised version of the manuscript.

2. replace ‘ocean’ with ‘sea’.

We corrected this.

3. 6, 33 ‘such as’ > ‘such that’.

We corrected this within the manuscript. We also changed this on Figs. 7 and 10.

4. 7, 13 Li et al (2013) is missing from the references

The reference has been added.

5. 7, 27-28 needs editing.

This sentence has been rephrased to "...This correction results to an improvement of the AOD product. Comparison against spatially and temporally co-located AERONET observations (Amiridis et al., 2013) returned an absolute bias of ~ -0.03..."

6. 8, section 2.4: give the definition of the AAI (I assume you mean the absorbing aerosol index) and explain that is a qualitative measure for aerosol absorption, give a reference.

We thank the reviewer for giving us the opportunity to clarify this. We mention in the revised version of the manuscript that this is the UV Aerosol Index (AI) also known as Absorbing Aerosol index and it is a qualitative indicator of the presence of UV absorbing aerosols in the atmosphere such as biomass burning and dust. We added the following reference: (Torres et al., 1998) and we also mention that positive AI values generally represent absorbing aerosols while small or negative values represent non-absorbing aerosols.

7. 8, 25 ‘is applied on spectral measurements from both TOMS and OMI sensor’ > ‘is applied to spectral measurements from both the TOMS and OMI sensors’.

We corrected this.

8. 14, 5 is there a reason for these large overestimations of sea salt contribution to the AOD?

Probably it has to do with the use of Smirnov (2003) formula for the calculation of marine AOD. It works well for open seas where we have stronger winds and sea salt dominates (when other types of aerosols dominate usually their AODs are pretty high, e.g. Saharan dust over the Atlantic Ocean). So, when the formula returns values higher than the MODIS AOD all the AOD is attributed to sea salt which is OK for
marine aerosol dominated regions. However, for the Mediterranean Basin (closed sea and not an open ocean) where other types of aerosols dominate the formula leads to a very frequent attribution of the total AOD to sea salt. Consequently this leads to a significant overestimation of marine aerosols. As there is already enough discussion on the tests we did in order to fine-tune the algorithm we prefer not to get into details about the reason of the overestimations.

We also have to mention that we mistakenly wrote in the original version of the paper that we used the 0.5 critical value for AI. We used this value in our tests but in the end we preferred to use 1 which is in accordance to other studies in the area. We have corrected Fig. 2 accordingly.

9. 17, 21 is it useful to give both the EE and pIEE? They are not very different and in principle do not provide additional info for this kind of study, or does it?

We believe yes. We wanted to show that even though in some cases the MODIS data have a lower percentage than 66% within the EE (the EE is a result of global validation studies) can still have a percentage within the pIEE which was defined encompassing the sum of absolute and relative AOD errors prior to the launch of Terra (Kaufman et al., 1997).

10. 18, 10 'about' > 'for the'.

We corrected this.

11. 18, 16 why don’t you use the same area as for DT?

Basically, we use the same spatial window for the validation (25 x 25 km²) for both DT and DB. In order to be in line with previous studies using Col. 5 MODIS AOD550 from DT and DB we also did a DT validation using the 50 km window as in Levy et al. (2010) and a DB validation using high quality retrievals and a 25 km window as in e.g. Shi et al. (2011). In the case of DB we also wanted to show that the use of all the data instead of high quality data only did not change the correlation with the ground-based data significantly but diminished the statistical sample by a factor of 5.

12. 18, 18 25x25.

We corrected this.

13. 19,22 taking this into account

We corrected this.

14. 20, 26 and elsewhere throughout the MS (e.g. 30, 22): formulation. Due to the low precipitation (removal) the AOD remains high, but actually the AOD is high because of emissions and atmospheric processes forming aerosol particles

We thank the reviewer for giving us the opportunity to correct this in the revised manuscript. We added the following in 20,26: "...It has to be highlighted that the AOD550 over these regions is high primarily due to the emissions and the
atmospheric processes forming aerosol particles. The low removal rates from precipitation just preserve the AOD$_{550}$ levels high..." and in 30,22: "...Precipitation is the major washout mechanism of atmospheric pollutants. Low removal rates from precipitation contribute in preserving high the AOD$_{550}$ levels which are a result of emissions and other atmospheric processes...

15. 21, 4 same as 20,26: the sources are what they are (important when large) but they do not become more prominent due to limited washout.

We rephrased the sentence to: "...The limited washout by precipitation (see also Papadimas et al., 2008) and also the enhanced photochemical production of secondary organic aerosols (Kanakidou et al., 2011 and references therein) contribute to the high AODs appearing over local sources..."


We corrected this.

17. 28, 33 remove ‘region of’.

We corrected this.

18. Figure 2: the text in the boxes is very small: could you make this larger?

We did them larger where possible (all the boxes except for the last line). We will keep the figure as big as possible in the final paper.

19. Figure 3: the headers above the text are difficult to read and not needed since they are all the same; removing them leaves more space to increase the font size of the statistical results given there. In the caption an expected uncertainty is mentioned (line 12) which does not occur in the text. Is this different from the EE, and if so, please explain how evaluated.

We agree with the reviewer, we have revised the Figure accordingly. The expected Uncertainty refers to the MODIS DB data and is given in Section 4.1 "... The MODIS Terra DB data overestimate AOD$_{550}$ by 21.38 % (NMB) with 51.90 % of the data falling within the expected uncertainty (EU) envelope assuming a DB expected uncertainty of ±0.05±20%AOD$_{AERONET}$ (Hsu et al., 2006)...."
from the AErosol RObotic NETwork (AERONET)

(Lelieveld et al., 2002)
the ocean
So, we can directly compare the MODIS AOD$_{550}$ retrievals against AERONET observations.
is applied

improved

improvement of the AOD\textsubscript{532} product. Comparison

AOD\textsubscript{532} absolute bias of \(~ -0.03\) compared against

to

returned an absolute bias of \(~ -0.03\)

UV (latitude)

( longitude)

( also known as Absorbing Aerosol index)

English (U.K.)

(e.g.

such as

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(Torres et al., 1998). Positive AI values generally represent absorbing aerosols while small or negative values represent non-absorbing aerosols.

(in this work 1°x1°)

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The results did not change significantly when using discussed in Jones and Christopher (2011),
which is suggested in Jones and Christopher, 2011) and therefore we decided to adopt 1 as the AI critical value.

is capable of separating sea salt from UV absorbing aerosols (dust and biomass burning) efficiently.

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contribute to the high AODs appearing over local sources

extending across the North African coast and the neighboring oceanic areas

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Summary and Conclusions

Precipitation as Low removal rates from precipitation contribute in preserving high the AOD$_{550}$ levels which are a result of emissions and other atmospheric processes.


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\degree x 0.1\degree gridded MODIS dataset was compiled and
validated against AERONET sunphotometric observations from the AErosol RObotic NETwork (AERONET). The high spatial resolution and long temporal coverage of the dataset allows for the determination of local hot spots like megacities, medium sized cities, industrial zones, and power plant complexes, seasonal variabilities, and decadal averages. The average AOD at 550 nm (AOD$_{550}$) for the entire region is $\sim 0.22 \pm 0.19$ with maximum values in summer and seasonal variabilities that can be attributed to precipitation, photochemical production of secondary organic aerosols, transport of pollution and smoke from biomass burning in Central and Eastern Europe, and transport of dust from the Sahara Desert and the Middle East. The MODIS data were analyzed together with data from other satellite sensors, reanalysis projects and a chemistry-aerosol-transport model using an optimized algorithm tailored for the region and capable of estimating the contribution of different aerosol types to the total AOD$_{550}$. The spatial and temporal variability of anthropogenic, dust and fine mode natural aerosols over land and anthropogenic, dust and marine aerosols over the sea is examined. The relative contribution of the different aerosol types to the total AOD$_{550}$ exhibits a low/high seasonal variability over land/oceanic sea areas, respectively. Overall, anthropogenic aerosols, dust and fine mode natural aerosols account for $\sim 51 \%$, $\sim 34 \%$ and $\sim 15 \%$ of the total AOD$_{550}$ over land, while, anthropogenic aerosols, dust and marine aerosols account $\sim 40 \%$, $\sim 34 \%$ and $\sim 26 \%$ of the total AOD$_{550}$ over the sea, based on MODIS Terra and Aqua observations.

1 Introduction

For more than fifteen years, two MODIS (Moderate Resolution Imaging Spectroradiometer) satellite sensors monitor tropospheric aerosols at a global scale on a daily basis. The retrieved aerosol optical properties have been used in numerous air quality studies as well as studies related to the effect of airborne particles on various climatic parameters (e.g. radiation, clouds, precipitation, etc.). The $1^\circ \times 1^\circ$ daily gridded level-3 dataset is primarily used for global as well as regional studies while the single pixel level-2 data with a 10 km resolution (at nadir) are mostly used for regional and local scale studies. Nevertheless, the use of the coarse resolution MODIS data has predominated even in regional studies. The reasons for this could be the smaller file size which makes their processing and storage easier or the fact that they are easily accessible through user-friendly data bases which also allow for a very basic analysis like e.g. NASA's GIOVANNI website (http://giovanni.gsfc.nasa.gov/giovanni/) (Acker and Leptoukh, 2007).
The same holds for studies focusing on the Mediterranean Basin, an area which is considered of particular sensitivity as far as air pollution and climate change is concerned (Lelieveld et al., 2002; Giorgi, 2006). The Mediterranean basin is one of the regions with highest aerosol optical depths (AODs) in the world (Husar et al., 1997; Ichoku et al., 2005; Papadimas et al., 2008), causing significant climate forcing especially in summer, which is characterized by low cloudiness and high incoming solar radiation levels (Papadimas et al., 2012; Alexandri et al., 2015). The Mediterranean is also recognized as a crossroads between three continents where aerosols of various types accumulate (Lelieveld et al., 2002). Marine aerosols from the Mediterranean Sea and even the Atlantic Ocean combine with aerosols from continental Europe (urban and rural), dust particles transported from the Sahara Desert and Middle East as well as biomass burning aerosols from occasional wild fires and agricultural burning (Lelieveld et al., 2002). Specifically, as discussed in Hatzianastassiou et al. (2009), the Eastern Mediterranean, the region under investigation here, is located at a “key” point of this crossroads. There is a significant number of ground and satellite-based studies on the abundance and optical properties of tropospheric aerosols in the area; however, these studies are either focused on specific spots or applied used a coarse spatial and temporal resolution.

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

On the other hand, AOD and other aerosol optical properties have been studied over the greater Eastern Mediterranean region based on data from Meteosat (Moulin et al., 1998), SeaWIFS (Koren et al., 2003; Antoine and Nobileau, 2006; Mélin et al., 2007; Nabat et al., 2013), TOMS (Alpert and Ganor, 2001; Israelevich et al., 2002; Koukouli et al. 2006; Hatzianastassiou et al., 2009; Koukouli et al., 2010, Israelevich et al., 2012, Kaskaoutis et al., 2012a; Nabat et al., 2013;
Gkikas et al., 2013–2014; Varga et al., 2014), MODIS Terra and Aqua (Barnaba and Gobbi, 2004; Papayannis et al., 2005; Kaskaoutis et al., 2007–2008–2010–2011–2012a,b,c,d; Kosmopoulos et al., 2008; Papadimas et al., 2008, 2009; Rudich et al., 2008; Carmona and Alpert, 2009; Karnieli et al., 2009; Gkikas et al., 2009–2013; Hatzianastassiou et al., 2009; El-Metwally et al., 2010; Koukouli et al., 2010; Kanakidou et al., 2011; Gerasopoulos et al., 2011; de Meij and Lelieveld, 2011; Marey et al., 2011; de Meij et al., 2012; Nabat et al., 2012, 2013; Nikitidou and Kazantzidis, 2013; Athanasiou et al., 2013; Benas et al., 2011–2013; Sorek-Hamer et al., 2013; Kabatas et al., 2014; Kourtidis et al., 2014; Mishra et al., 2014; Flaounas et al., 2015; Kloog et al., 2015), OMI/AURA (Kaskaoutis et al., 2010; El-Metwally et al., 2010; Marey et al., 2011; Kaskaoutis et al., 2012b, c; de Meij et al., 2012; Nabat et al., 2012, 2013; Nikitidou and Kazantzidis, 2013; Athanasiou et al., 2013; Benas et al., 2011–2013; Sorek-Hamer et al., 2013; Varga et al., 2014; Flaounas et al., 2015), CALIOP/CALIPSO (Amiridis et al., 2009, 2013, Mamouri et al., 2009; Marey et al., 2011; Kaskaoutis et al., 2012c; de Meij et al., 2012; Nabat et al., 2012, 2013; Mamouri and Ansmann, 2015), MISR/Terra (Kanakidou et al., 2011; Marey et al., 2011; de Meij and Lelieveld, 2011; de Meij et al., 2012; Nabat et al., 2013; Kabatas et al., 2014; Abdelkader et al., 2015) as well as NOAA/AVHRR, MERIS/ENVISAT, AATSR/ENVISAT, PARASOL/POLDER, MSG/SEVIRI, and Landsat satellite data (see Retalis and Sifakis, 2010; Nabat et al., 2013; Benas et al., 2013; Sifakis et al., 2014). To our knowledge, these studies comprise the majority of work focusing on tropospheric aerosols over the Eastern Mediterranean by means of satellite remote sensing, published in peer reviewed journals the last ~ 15 years. As shown in Fig. 4 of this work, the publication rate of satellite-based studies focusing on the Eastern Mediterranean aerosols nearly doubled every three years during the period 1997-2014 which is indicative of the increasing scientific interest in the area.

In a very large fraction of the satellite-based studies referenced above, the used data are either of coarse mode (usually 1° which is ~ 100 km for the mid-latitudes) or focus on specific spots for validation purposes. In a few cases, high resolution data were used in spatiotemporal studies; however, either these studies are restricted over surfaces covered by water or examine a short period only. For example, Moulin et al. (1998) investigated the dust AOD patterns over the oceanic areas of the Mediterranean Basin at a resolution of 35 x 35 km² for a period of 11 years (1984-1994) using Meteosat observations. A 7-year climatology (1998-2004) of total and dust AOD for the same regions at a resolution of 0.16° x 0.16° was compiled by Antoine and Nobileau (2006) using observations from SeaWIFS. Mélin et al. (2007) merged SeaWIFS and MODIS data and presented high resolution AOD patterns (2 x 2 km²) for May 2003. As far as MODIS is concerned, only Barnaba and Gobbi (2004) presented a high resolution (0.1° x 0.1°)
spatiotemporal analysis for a period of 1 year (2001) over ocean only. In a recent paper, Athanasiou et al. (2013) presented in detail a method for compiling a 0.5-degree resolution AOD gridded dataset using level-2 MODIS Terra data for the greater region of Greece (2000-2008). However, the spatial resolution they used (~ 50 km) is not high enough to reveal local sources (e.g. cities, islands, river banks, etc.). Overall, there has not been so far any detailed high resolution spatiotemporal study of the AOD over the Eastern Mediterranean. In this paper, the AOD_{550} spatiotemporal variability over the Eastern Mediterranean (30°N-45°N, 17.5°E-37.5°E) is presented at a high spatial resolution (0.1° x 0.1°) based on MODIS Terra and Aqua observations. Level-2 MODIS data are used for the compilation of a 0.1-degree gridded dataset which is validated against ground-based observations. In order to calculate the contribution of different aerosol types to the total AOD, the MODIS data were analyzed together with other satellite data, ERA-Interim reanalysis data and the Goddard Chemistry Aerosol Radiation and Transport (GOCART) model using an algorithm optimized for the surface properties of the Eastern Mediterranean region. The different datasets used in this research are presented in detail in Sect. 2 while a detailed description of the method is given in Sect. 3. Sect. 4 includes the results from the MODIS validation procedure, the annual and seasonal variability of AOD_{550} over the region with a discussion on the local aerosol sources and the differences between Terra and Aqua, and the annual and seasonal contribution of different aerosol types to the total AOD_{550}. Finally, in Sect. 5, the main conclusions of the paper are presented along with a short discussion on how these results could contribute to future studies in the area.

2 Observations, reanalysis data and model simulations

2.1 MODIS Terra and Aqua satellite observations

The main data used in this work come from the level-2 MODIS Terra (MOD04_L2) and MODIS Aqua (MYD04_L2) Collection 051 dataset and have been acquired through NASA’s Level 1 and Atmosphere Archive and Distribution System (LAADS) (http://ladsweb.nascom.nasa.gov). The fact that MODIS Terra and Aqua have a daytime equator crossing time at 10:30 LT (morning) and 13:30 LT (noon), respectively. MODIS instruments with a viewing swath of 2330 km measure backscattered radiation at 36 spectral bands between 0.415 and 14.235 μm with a spatial resolution of 250, 500 and 1000 m, providing a nearly global coverage on a daily basis. Aerosol optical properties for the standard MODIS aerosol product are retrieved using two different "Dark Target" (DT)
algorithms. The one is used over land surfaces (Kaufman et al., 1997; Remer et al., 2005; Levy et al., 2010) and the other over oceanic regions (Tanré et al., 1997; Levy et al., 2003; Remer et al., 2005). The "Deep Blue" algorithm (DB) (Hsu et al., 2004; Hsu et al., 2006) has been used for retrievals over bright land surfaces (e.g. deserts) where the DT algorithm fails. Only recently, updates to the algorithm allowed for extending the spatial coverage of the DB aerosol product over all land areas (Hsu et al., 2013; Levy et al., 2010; Hsu et al., 2006). AERONET Cimel sunphotometric measurements have been extensively used for the validation of the MODIS over-land and over-ocean products (e.g. Chu et al., 2002; Remer et al., 2002; Remer et al., 2005; Levy et al., 2010; Shi et al., 2013).

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

### 2.2 AERONET ground-based observations

For the evaluation of the MODIS AOD$_{550}$, data from 13 AERONET Cimel network ground stations in the region of the Eastern Mediterranean have been acquired (http://aeronet.gsfc.nasa.gov). The stations were selected such as that their operation period...
covers at least 2 years and there are at least 100 common days of co-localized AERONET and MODIS observations. AERONET Cimel sunphotometers measure solar radiation every 15 minutes within the spectral range from 340 to 1020 nm (Holben et al., 2001). The spectral measurements allow for the retrieval of columnar aerosol properties (see Holben et al., 1998; Dubovik and King, 2000; Dubovik et al., 2000, 2002). The AERONET AOD uncertainty is in the order of 0.01–0.02 (Eck et al., 1999), being larger at shorter wavelengths. Here, we use quadratic fits on a log-log scale to interpolate the AERONET data (AODs at 440, 500, 675 and 870 nm) to the MODIS band-effective wavelength of 550 nm (Eck et al., 1999; Levy et al., 2010). So, we can directly compare the MODIS AOD_550 retrievals against AERONET observations. Simultaneous measurements of the Ångström Exponent (AE) for the spectral range 440–870 nm (AE_{440,870}) from the 13 AERONET stations mentioned above were also utilized in this work in order to account for days with dust dominance. The uncertainty of the AE is significantly higher than the AOD uncertainty, especially under low-AOD conditions. Li et al. (2014) found that the uncertainty for a typical Northern Hemispheric AERONET station (GSFC) is ~ 0.6 during winter when AODs are significantly lower compared to summer (~ 0.15).

2.3 LIVAS CALIOP/CALIPSO dust climatology

Dust aerosol optical depths at 532 nm (AOD_{532}) from CALIOP/CALIPSO (Cloud-Aerosol Lidar with Orthogonal Polarization) satellite at a resolution of 1° x 1° are also used here for the period 2007-2012. CALIPSO measures cloud and aerosol properties flying at a 705 km sun synchronous polar orbit with a 16 day repeat cycle and an equator-crossing time close to that of the Aqua satellite (13:30 LT). The dust product used here comes from a Saharan-dust-optimized retrieval scheme that was developed within the framework of the LIVAS (Lidar Climatology of Vertical Aerosol Structure for Space-Based LIDAR Simulation Studies) project (Amiridis et al. 2015) and has been presented in detail in Amiridis et al. (2013). In brief, the LIVAS dust product is optimized for Europe by applying a lidar ratio of 58 sr instead of 40 sr to Level 2 dust related backscatter products. This correction results to an improved improvement of the AOD_{532} product. Comparison AOD_{532} absolute bias of ~ -0.03 compared against spatially and temporally co-located AERONET observations returned an absolute bias of ~ -0.03. The corresponding reported biases for the original CALIPSO data are significantly higher (~ -0.10). The bias is even lower when
compared against MODIS satellite-based observations. Other improvements of this product are related to the use of a new methodology for the calculation of pure dust extinction from dust mixtures and the application of an averaging scheme that includes zero extinction values for the non-dust aerosol types detected. Overall, this product (hereafter denoted as LIVAS dust product) exhibits better agreement with observations from MODIS and AERONET and simulations from the BSC-DREAM8b dust model over North Africa and Europe than the standard CALIPSO data hence being an ideal tool for the evaluation of other satellite-based products.

2.4 Earth Probe TOMS and OMI satellite observations

For this work, UV Aerosol Index (AI) data (Herman et al., 1997) from the Earth Probe TOMS (Total Ozone Mapping) spectrometer aboard Earth Probe for the period 1/2000-9/2004 at a resolution of 1° (latitude) x 1.25° (longitude) and the OMI (Ozone Monitoring Instrument) sensor aboard EOS AURA for the period 10/2004-12/2012 at a resolution of 1° x 1° were acquired through the GIOVANNI web database (http://giovanni.gsfc.nasa.gov/giovanni/).

Earth Probe TOMS continued the record of the first three TOMS instruments aboard Nimbus-7, Meteor-3 and ADEOS flying in a sun synchronous orbit at an altitude of 740 km with an instantaneous field of view size of 39 x 39 km² at nadir. The instrument had an ascending node equator crossing time at 12:00 LT covering 85 % of the globe on a daily basis from 7/1996 until 12/2005. The satellite was originally set to a 500 km sun synchronous orbit but was set to its final orbit after the failure of ADEOS satellite in 6/1997. OMI is a UV/VIS nadir solar backscatter spectrometer (Levelt et al., 2006) that continues the long TOMS record. OMI flies in a sun synchronous polar orbit at an altitude of 705 km with an ascending node equator crossing time at 13:45 LT. Its 2600 km viewing swath allows for almost daily global coverage while the spatial resolution of the instrument is 13 x 24 km² at nadir. The AI (also known as Absorbing Aerosol index) which is calculated by the two instruments constitutes a qualitative indicator of the presence of UV absorbing aerosols in the atmosphere (e.g. such as biomass burning, and dust) (Torres et al., 1998). Positive AI values generally represent absorbing aerosols while small or negative values represent non-absorbing aerosols.

The Version 8 algorithm is applied on-to spectral measurements from both TOMS and OMI sensor to produce a consistent long-term AI timeseries (Li et al., 2009). AI is calculated from the difference in surface reflectivity derived from the 331.2 and 360 nm measurements exhibiting an uncertainty of ±30 % (Torres et al., 2007).
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 09:00 and 12:00 UTC on a daily basis for the period 2000-2012. We use 09:00 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 ~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°x1°) 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$_{550}$ 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$_{550}$ 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 ~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.,...
showing that the MACC aerosol products generally capture well the daily, seasonal and interannual variability of aerosols. As discussed in Bellouin et al. (2013) the uncertainties of MACC total AOD$_{550}$ (~ 0.03) and dust AOD$_{550}$ (~ 0.014) arise from uncertainties in the MODIS retrievals which are assimilated into the model and errors in the forward modeling of total and component AODs.

2.7 GOCART data

Daily total and dust AOD$_{550}$ data from the GOCART chemistry-aerosol-transport model simulations (version 006) are used in this study for the period 2000-2007. The GOCART model (see Chin et al., 2000, 2002, 2004, 2007; Ginoux et al., 2001, 2004) uses the assimilated meteorological fields of the Goddard Earth Observing System Data Assimilation System (GEOS DAS) which are generated by the Goddard Global Modeling and Assimilation Office (GMAO). The data which are used were acquired from an older version of NASA's GIOVANNI web database (http://disc.sci.gsfc.nasa.gov/giovanni/) and come from a simulation implemented at a spatial resolution of 2° (latitude) x 2.5° (longitude) with 30 vertical sigma layers (Chin et al., 2009). The model includes physicochemical processes of major tropospheric aerosol components (sulfates, dust, black carbon, organic carbon, sea salt) and precursor gases (SO$_2$ and dimethylsulfide) incorporating various atmospheric processes. The total AOD$_{550}$ from GOCART compared to ground-based observations from the AERONET exhibits a relative mean bias [mean(GOCART)/mean(AERONET)] of 1.120, 1.135 and 0.959 over Europe, North Africa and for the whole globe, respectively.

2.8 Ancillary data

Apart from the primary datasets presented above, three additional datasets were used in order to support our findings. OMI/AURA daily gridded (Bucsela et al., 2013) tropospheric NO$_2$ columnar data (OMNO2d version 2.1) at a spatial resolution of 0.25° x 0.25° were acquired from NASA's GIOVANNI web database (http://giovanni.gsfc.nasa.gov/giovanni/) for the period 2005-2012. The quality checked data used in this work correspond to sky conditions where cloud fraction is less than 30 %. Planetary boundary layer (PBL) SO$_2$ daily gridded columnar data (OMSO2e version 1.1.7) were also acquired from GIOVANNI for the same period. The OMSO2e gridded data (0.25° x 0.25°) used in this work are produced from best level-2 pixel data, screened for OMI row anomaly and other data quality flags. The PBL SO$_2$ column retrievals are produced with an algorithm based on principal component analysis
(PCA) of the OMI radiance data (Li et al., 2013). Finally, monthly precipitation data from the 3B43 TRMM and Other Sources Monthly Rainfall Product (version 7) at a spatial resolution of 0.25° x 0.25° for the period 2000-2012 were obtained from GIOVANNI. This dataset is derived from 3-hourly precipitation retrievals from the Precipitation Radar (PR), the TRMM Microwave Imager (TMI) and the Visible and Infrared Scanner (VIRS) aboard the TRMM (Tropical Rainfall Monitoring Mission) satellite merged with other satellite-based precipitation data and the Global Precipitation Climatology Centre (GPCC) rain gauge analysis (Huffman et al., 2007).

3 Methodology

3.1 Compiling a MODIS 0.1° x 0.1° gridded dataset

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

First, a 0.1° x 0.1° resolution grid covering the Eastern Mediterranean (30°N-45°N, 17.5°E-37.5°E) is defined which corresponds to 30000 grid cells. As already mentioned in Sect. 2.1, only level-2 single pixel AOD$_{550}$ measurements with a QAC flag of 3 and a QAC flag greater than 0 were used over land and over ocean, respectively, to ensure the high quality of the data. Pixels are attributed to a specific grid cell if their center falls within a 25 x 25 km$^2$ square window around the grid cell (see Fig. S1 in the Supplement). These pixels are then used for the calculation of daily averages. As shown in Figure S1, a grid cell of 0.1° (~ 10 km) is as big as the centre of a large Mediterranean city like Thessaloniki, Northern Greece (~ 1 million inhabitants). The procedure was followed separately for MODIS Terra and Aqua data. In cases of grid cells with no DT MODIS observations, data from the DB algorithm were used (over
The size of the gridding window was selected following Koukouli et al. (2007). They used both 10 and 25-km windows showing that the latter allows for the inclusion of more data points without undermining the ability of monitoring accurately the aerosol load over a specific spot. In addition, in cases of urban sites, a window of 25 km allows for the inclusion of pixels from the surrounding non-urban surfaces where the MODIS surface reflectance parameterization is better (Levy et al., 2010). The size of each MODIS pixel is 10 km at nadir, but at the swath edges, it may become 2-3 times larger. Hence, ideally the maximum number of pixels that could be used in the daily averaging is nine. The overlap between the windows of neighbouring grid cells does not affect the representativeness of the dataset over each grid cell. Aerosols are transported by air masses throughout the day and thus the aerosol load in neighbouring grid cells is not expected to be completely independent.

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

In order to generalize our results, nine different sub-regions (Fig. 1) were selected apart from the three basic regions of interest, namely, the whole Eastern Mediterranean (EMT) and the land (EML) and oceanic (EMO) areas of the region. The selection was done mainly taking into account geographical but also land type and land use criteria. The four sub-regions that correspond to the land regions of the Eastern Mediterranean are the Northern Balkans Land...
(NBL), the Southern Balkans Land (SBL), the Anatolia Land (ANL) and the Northern Africa Land (NAL) region while the five sub-regions that correspond to the oceanic regions are the Black Sea Oceanic (BSO), the North-Western Oceanic (NWO), the South-Western Oceanic (SWO), the North-Eastern Oceanic (NEO) and the South-Eastern Oceanic (SEO) region. Mean values of the total AOD$_{550}$ from the Terra and Aqua MODIS are reported for each one of the three basic regions of interest and their nine sub-regions in Sect. 4.

3.2 Contribution of different aerosol types to AOD$_{550}$

3.2.1 OceanSea

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

In general, the algorithm used over the oceanic regions (see Fig. 2) is similar with the one presented in Bellouin et al. (2008). First, the marine AOD$_{550}$ ($\tau_m$) is calculated from near surface wind speed using a linear relation which has been obtained from ground-based studies over pollution free oceanic regions. Bellouin et al. (2008) use the linear relation of Smirnov (2003). Then, if $\tau_m$ is greater or equal than AOD$_{550}$ it is assumed that there are marine particles only over this region. If $\tau_m$ is smaller than AOD$_{550}$ a decision tree is followed which is first based on FMR$_{550}$ and then on AI in order to reach a conclusion about the type of aerosols that account for AOD$_{550}$. If FMR$_{550}$ is smaller than the critical value of 0.35 and AI is greater than or equal to a critical value it is assumed that there are both marine aerosols ($\tau_m$) and dust ($\tau_d$=AOD$_{550}$-$\tau_m$) while if AI is smaller than this critical value it is assumed that there are marine aerosols only.

The AI critical value is equal to 1 in Bellouin et al. (2008). If FMR$_{550}$ is greater than or equal to 0.83 it is assumed that there are both anthropogenic ($\tau_a$=AOD$_{550}$-$\tau_m$) and marine aerosols ($\tau_m$). In the occasion of having a FMR$_{550}$ equal to 0.35 or greater than 0.35 but smaller than 0.83 one has to take again AI into consideration. If AI is less than the critical value it is assumed that there are marine aerosols ($\tau_m$) only while in the opposite occasion it is assumed that all the three
types of aerosols that can be defined over oceanic regions by this algorithm, namely, dust
\[ \tau_d = (1 - FMR_{550})(AOD_{550} - \tau_m) \], anthropogenic \[ \tau_a = FMR_{550}(AOD_{550} - \tau_m) \] and marine aerosols \( \tau_m \) are present. One should keep in mind that all the biomass burning aerosols are classified as anthropogenic by this method.

In this work, we proceeded to a "fine-tuning" of the algorithm for the region of the Eastern Mediterranean. First, we applied the algorithm on MODIS Terra data using the same critical equations and critical values as in Bellouin et al. (2008). The results showed that the original Bellouin et al. (2008) method might be valid for global studies but for a "closed" sea like the Mediterranean the use of the original critical values leads to a large overestimation of sea salt AODs and therefore underestimation of dust and anthropogenic aerosol AODs. Indicative of this situation is Fig. S3 in the Supplement where we present the relative contribution of dust, marine and anthropogenic aerosols per month over the oceanic regions of the Eastern Mediterranean as calculated using the original Bellouin et al. (2008) method. It is shown that the marine contribution is several times higher than the values reported for the Mediterranean Basin in previous studies (e.g. see Nabat et al., 2012). Evaluation of the algorithm was done using dust AOD_{532} data from the LIVAS CALIOP/CALIPSO product. From LIVAS we only use the high quality Sahara dust product as a reference and not other aerosol type retrievals (e.g. marine aerosols) since the dust retrievals from CALIOP/CALIPSO are by far the most reliable (e.g. Burton et al., 2013). We performed several tests by changing the linear relation that connects \( \tau_m \) with near surface wind speed and the AI critical values and compared each time the dust AOD_{550} seasonal variability with the LIVAS AOD_{532} seasonal variability for the ocean sea covered sub-regions of the Eastern Mediterranean. Results from this algorithm-tuning procedure can be found in Figs. S4e-i of the Supplement where one can also see the underestimation of dust AOD_{550} from the original Bellouin et al. (2008) algorithm. The linear relation given in Kaufman et al. (2005) was finally selected \( \tau_m = 0.007ws + 0.02 \). The 2000-2012 average wind speed over the ocean sea for the region of the Eastern Mediterranean is \(~ 5.3 \text{ m/s}. \) Kaufman et al. (2005) reduced the offset in the linear relation of Smirnov (2003) from 0.06 to 0.02 to fit the average baseline AOD of 0.06 for the typical wind speed of 6 m/s. In addition, our tests showed that an AI critical value of 0.51 performs better over the region of the Eastern Mediterranean. As the results did not change significantly when using discussed in Jones and Christopher (2011), other AI thresholds of \( \text{e.g. } 0.5 \text{ which is suggested in Jones and Christopher, 2011} \) and therefore we decided to adopt 1 as the AI critical value is capable of separating sea salt from UV absorbing aerosols (dust and biomass burning) efficiently. Another
test, following the example of other studies (see Lehahn et al., 2010), was to assume that for wind speed less than 5 m/s there is very little or no sea-spray particle production (limited bursting of entrained air bubbles associated with whitecap formation). In this case, $\tau_{in}$ is stable, equal to the offset of the linear relation between $\tau_{in}$ and wind speed which is indicative of the background sea salt AOD$_{550}$. However, this test reveals that the effect of assuming stable $\tau_{in}$ for wind speed less than 5 m/s is insignificant and therefore we selected to follow the Kaufman et al. (2005) linear relationship for the whole wind speed range. As shown in Figs. S4e-i, the seasonal variability when applying our modified algorithm over oceanic regions is very close to the LIVAS dust AOD$_{552}$ especially for the months with lower dust load (June-January). It is also shown that dust AODs from this algorithm are closer to the LIVAS dust product than dust AODs from MACC reanalysis do. The slight overestimation of dust AOD or the shift of the maximum dust load we observe for the period of high dust loads in the region (February-May) is probably connected to the narrow swath and the 16-day time of CALIPSO which means that several dust events might be not observed by the CALIOP instrument contrary to MODIS which has a daily coverage.

### 3.2.2 Land

As already mentioned in the previous paragraph a different approach is followed over the land regions of the Eastern Mediterranean due the low confidence on the MODIS FMR$_{550}$ and Ångström exponent retrievals over land compared to that over ocean (e.g. see Levy et al., 2010; Georgoulas and Kourtidis, 2011). This limitation does not allow us to distinguish the contribution of fine and coarse mode aerosols in terms of AOD$_{550}$. In this case, we choose to use daily model fields of the dust contribution to the total AOD (here MACC reanalysis and GOCART). We follow a method similar with the one presented in Bellouin et al. (2013). Specifically, we calculate the dust AOD$_{550}$ by scaling the MODIS AOD$_{550}$ data with the MACC or GOCART dust/total AOD$_{550}$ ratios $[f_d = \tau_d/(\tau_{total})]$ on a daily basis. Since the MACC data are available only from 2003 to 2012, in order to take advantage of the full MODIS dataset (3/2000-12/2012), data from the GOCART model were used for the period 2000-2002. The GOCART data were normalized in order to be consistent with the MACC data. Daily dust/total AOD$_{550}$ ratios ($f_d$) from the common GOCART-MACC period 2003-2007 were first brought to a common 1° x 1° spatial resolution using bilinear interpolation and then we calculated the regression line for each grid cell on a seasonal basis. The linear relations were afterwards used in order to normalize the 2000-2002 GOCART ratios to have a homogeneous
dataset. The slopes and offsets of these regression lines and the corresponding correlation coefficients (R) can be seen in Figs. S5, S6 and S7 of the Supplement, respectively. Overall, for the whole time period, the MACC reanalysis \( f_d \) ratios are lower by ~ 26 % from the GOCART \( f_d \) ratios and the linear relation connecting the two products is \( f_{d\text{MACC}}=0.4964f_{d\text{GOCART}}+0.0952 \) with a correlation coefficient \( R \) of 0.74. The \( f_d \) values of the merged GOCART-MACC (2000-2012) timeseries were checked using the Standard Normalized Homogeneity Test (SNHT) as described in Alexandersson (1986). The statistical significance was checked following Khaliq and Ouarda (2007) and the \( f_d \) timeseries were found to be homogeneous (see Fig. S8 of the Supplement). Hence, this test verifies that the use of the merged GOCART-MACC \( f_d \) dataset will not insert any artifacts (e.g. trends or breaks) in the algorithm. Finally, the \( f_d \) data were brought to the same spatial resolution with MODIS data (0.1° x0.1°) using bilinear interpolation again.

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

Overall, the algorithm described above performs well as far as dust is concerned. This is further shown when comparing MODIS Terra and Aqua \( \tau_d \) values with collocated AERONET observations for dust dominated days (see Fig. S9 in the Supplement). The method followed for the collocation of the data is similar to the one presented in Sect. 4.1 while dust dominated days were days with an AERONET AE smaller than 1 (see Mateos et al., 2014) and a MODIS based \( \tau_d \) greater than \( \tau_n \) or \( \tau_m \). The uncertainties of the calculated \( \tau_a, \tau_d, \tau_n \) and \( \tau_m \) values which are inserted by the input data and the assumptions of the algorithm are expected to be similar with the ones presented in Bellouin et al. (2013). Bellouin et al. (2013) using a Monte-Carlo analysis indicated that \( \tau_a \) can be specified with an uncertainty of ~ 23 % over land and ~ 16 % over the ocean, \( \tau_d \) can be specified with an uncertainty of ~ 19 % over land and ~ 33 % over the
ocean, \( \tau_0 \) can be specified with an uncertainty of \( \sim 41\% \) and \( \tau_m \) with an uncertainty of \( \sim 28\% \).

The results of the application of the algorithm described in the paragraphs above are presented in the following section (Sect. 4) by means of maps, pie charts, plots and tables for each one of the three basic regions of interest and their nine sub-regions.

4 Results and discussion

4.1 Validation of MODIS gridded data using ground-based observations

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

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

To be in line with the global validation of the DT Collection 051 product by Levy et al. (2010) we also performed a validation with the specifications used in their work. We used a 50 x 50 km$^2$ window for the spatial collocation of the MODIS and AERONET data while only days with at least 5 MODIS retrievals and 2 AERONET measurements were taken into account. The increased size of the collocation window improves the results of the validation. As shown in Table 2, MODIS Terra DT Collection 051 data overestimate AOD$_{550}$ by 5.10 % (NMB) with 70.17 % of the data falling within the EE envelope and 74.64 % within the pEE envelope. For MODIS Aqua, the NMB is 15.34%, while the percentage of the measurements falling within the EE and pEE envelope is 66.76 % and 70.45 %, respectively. These results about for the Eastern Mediterranean are close to the global ones presented in Levy et al. (2010).

As discussed in Sect. 3.1, data from the DB algorithm were used over bright arid and semi-arid regions of North Africa for the production of the 0.1° x 0.1° MODIS gridded dataset for grid cells with no DT data. Therefore, in this work we also perform a validation of the DB Collection 051 product over the region of the Eastern Mediterranean. In the case of DB data, we first make use of all the available DB observations without any quality filtering over the 13 AERONET stations. A spatial window of 25-30 km has been typically used in the past for the collocation of MODIS DB data with the AERONET observations (see Shi et al., 2011; Ginoux et al.; 2012; Sayer et al., 2013; 2014) which is in line with the 0.25 x 0.25 km$^2$ window used here. The MODIS Terra DB data overestimate AOD$_{550}$ by 21.38 % (NMB) with 51.90 % of the data falling within the expected uncertainty (EU) envelope assuming a DB expected uncertainty of ±0.05 ± 20%AOD$_{AERONET}$ (Hsu et al., 2006). The MODIS Aqua DB Collection 051 data overestimate AOD$_{550}$ by 33.03 % (NMB) with 55.30 % of the data falling within the expected uncertainty envelope. We repeated the validation procedure for DB data taking into account the highest quality data only. The sample of available measurements was diminished by a factor of 5 in the case of MODIS Terra and 6 in the case of MODIS Aqua but the results were pretty similar with the ones for the unfiltered data. Therefore, the use of unfiltered DB data during the gridding procedure does not insert any significant uncertainty. The DB results for the 13 AERONET stations examined here are not of the same agreement.
than the DT results and the ones presented in previous studies utilizing DB Collection 051 data for other stations and larger regions (see Shi et al., 2011; Ginoux et al., 2012). However, it has been reported that stations in the region (e.g. Sede Boqer in Israel) are among the ones with the greatest discrepancies between MODIS DB and AERONET measurements (Ginoux et al., 2012). Nevertheless, as commented in Sect. 3.1, the DB data constitute only a small fraction of the data used for the production of the MODIS gridded dataset (~ 1 % only of the 30000 grid cells covering the Eastern Mediterranean has only DB retrievals) and therefore they do not affect significantly its quality.

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

The AOD\textsubscript{550} spatial variability over the greater Eastern Mediterranean region for the period 2000-2012 as seen from the Terra MODIS 0.1° x 0.1° dataset is presented in Fig. 4. Several aerosol hot spots that coincide with megacities (e.g. Cairo, Istanbul), large cities (e.g. Athens, Ankara, Alexandria, Izmir, Thessaloniki) or even medium sized cities (e.g. Larissa, Limassol), industrial zones (e.g. OSTIM Industrial Zone in Ankara, Turkey), power plant complexes (e.g. Maritsa Iztok complex at the Stara Zagora Province in Bulgaria, Ptolemaida-Kozani power plants in Western Macedonia, Greece), river basins (e.g. Evros river Basin at the borders between Greece and Turkey), etc, can be detected on the map. Indicatively, in Fig. 4 we give a list of 35 local particle pollution sources in the region; however, careful inspection of this map and the seasonal maps presented in Fig. 6 allows for the detection of many more aerosol sources. The results from the analysis of Aqua MODIS data are pretty similar as shown in Fig. S10 of the Supplement. A significant number of the local aerosol sources can also be detected on the OMI 2004-2012 tropospheric NO\textsubscript{2} and PBL SO\textsubscript{2} maps given in Figs. 5a and b which reveals the origin of aerosols over these regions (e.g. traffic, industrial activities, etc). However, there are regions of high aerosol load which cannot be seen in Fig. 5a and b and vice versa which is indicative of the significant role of other anthropogenic or natural processes that contribute to the local aerosol load (e.g. fires, soil dust from agricultural activities or arid regions, Sahara dust transport).

The topography (Fig. 5c) and precipitation (see Fig. 5d for annual precipitation levels for the period 2000-2012 from TRMM) are also major determinants of the local AOD\textsubscript{550} levels. For example, regions with mountain ranges on the Balkan Peninsula (e.g. Pindus mountain range in Greece, Dinaric Alps that run through Albania and the former Yugoslav republics, the Balkan mountain range in Central Bulgaria) are characterized by low AODs (see Fig. 4). On the contrary, regions of low altitude are generally characterized by higher AODs because the majority of anthropogenic activities is usually concentrated there. Also, low altitude regions surrounded by high mountains are characterized by higher AODs as aerosols cannot be easily transported by the wind (e.g. the industrialized regions in Central Bulgaria which are confined between the high Balkan and Rodopi mountain ranges). As precipitation is the major removal mechanism of pollutants in the atmosphere, regions with high AOD\textsubscript{550} are in many cases connected to low precipitation levels and vice versa (see Figs. 4 and 5d). It has to be highlighted that the AOD\textsubscript{550} over these regions is high primarily due to the emissions and the atmospheric processes forming aerosol particles. The low removal rates from precipitation
just preserve the AOD$_{550}$ levels high. A striking example is the one region of Anatolia in Central Turkey which is characterized by lower precipitation levels and higher aerosol loads compared to the surrounding regions. Also, the low precipitation levels are partly responsible for the high aerosol loads appearing over Northern Africa.

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

The AOD$_{550}$ spatial variability on a seasonal basis from MODIS Terra and Aqua is presented in Fig. 6 along with the difference between the two products. The majority of the local aerosol sources over land are more prominent in summer due to the limited washout by precipitation (see also Papadimas et al., 2008) and also due to the enhanced photochemical production of secondary organic aerosols (Kanakidou et al., 2011 and references therein) contribute to the high AODs appearing over local sources. In addition, during summer, over the region, there is typically a significant transport of aerosols (e.g. see Kanakidou et al., 2011 and references therein) and gaseous pollutants like SO$_2$ and NO$_2$ (see Georgoulias et al., 2009; Zyrichidou et al., 2009) and biomass burning aerosols from Central-Eastern Europe. Over the ocean sea, a profound maximum is observed in spring extending across the North African coast and the neighboring oceanic areas which is due to the well documented transport of significant amounts of dust from the Sahara Desert extending across the North African coast and the ocean see Barnaba and Gobbi., 2004 and the list of references given in the introduction). The seasonal variability of aerosols and the relative role of different aerosol types and various processes is discussed in more details in Sect 4.4.

The difference between MODIS Terra and Aqua Collection 051 AOD$_{550}$ over the Eastern Mediterranean is -0.002 (-1.40 %) for winter, -0.009 (-3.27 %) for spring, -0.011 (-4.46 %) for summer and 0.008 (4.40 %) for autumn. AOD$_{550}$ levels from Terra MODIS are lower than that from Aqua MODIS over land for all seasons. Over the sea, Terra MODIS AOD$_{550}$ levels are lower than that of Aqua MODIS only in winter. The fact that Terra MODIS measurements are systematically higher than that from Aqua over the ocean sea by ~ 0.01 on an annual basis is in line with the findings of previous global studies for Collection 5 (e.g. Remer et al., 2006;
Locally, one can see regions with positive and negative differences between Terra and Aqua MODIS AOD$_{550}$. The patterns of the Terra-Aqua difference per season are presented in Figs. 6c, f, i and l while the patterns of the percent difference are given in Fig. S11 of the Supplement. The largest part of the Terra-Aqua MODIS differences over land and ocean observed here may be attributed to the known calibration and sensor degradation issues of MODIS (for details see Levy et al., 2010; 2013; Lyapustin et al., 2014). A significant effort has been undertaken to address these issues in the new (Collection 6) MODIS product (e.g. Levy et al., 2013; Lyapustin et al., 2014) and a repetition of a similar analysis with Collection 6 data in the future would be a valuable contribution. Taking into account the aforementioned issues and the retrieval uncertainty of MODIS it becomes more than obvious that the attribution of observed differences between Terra and Aqua to the diurnal variability of aerosol load (e.g. over biomass burning regions) in the region is a difficult task. It is shown in Fig. S12 of the Supplement that the diurnal variability of AOD$_{550}$ from AERONET ranges significantly from station to station. The average hourly departure from the daily mean for the total of the 13 stations ranges from $\sim -5\%$ to $\sim 5\%$. Specifically, for the MODIS Terra and Aqua overpass times, the AERONET AOD$_{550}$ difference ranges from $\sim -10\%$ to $\sim 10\%$ (see Fig. S12b). The Terra-Aqua AOD$_{550}$ difference is negative for the total of the 13 stations ranging from $\sim -25\%$ to $\sim -5\%$. It is shown in Fig. S12b that the two differences exhibit a similar variability from station to station which indicates that part of the observed Terra-Aqua difference is indeed due to the diurnal variability of aerosols. However, as mentioned above, the diurnal variability of aerosols is a very delicate issue and should be comprehensively addressed in a future study. The same stands for other kind of variabilities which could be connected to local and regional anthropogenic activities like e.g. the weekly cycle of aerosols (see Georgoulias and Kourtidis, 2011; Georgoulias et al., 2015).

### 4.3 Contribution of different aerosol types to the total AOD$_{550}$

#### 4.3.1 Annual contribution

As mentioned above, we attempt to estimate in our work the contribution of different aerosol types to the total AOD$_{550}$ over the region of the Eastern Mediterranean was calculated following the methodology presented in Sect. 3.2. For the land covered areas, based on MODIS Terra observations, we estimate that 52% (0.112±0.087) of the total AOD$_{550}$ is due to anthropogenic aerosols, 32% (0.074±0.080) due to dust and 16% (0.034±0.026) due to fine mode natural aerosols (see Fig. 7). For the oceanic areas, 41% (0.086±0.085) of the total
AOD$_{550}$ is due to anthropogenic aerosols, 34 % (0.076±0.185) due to dust and 25 % (0.054±0.018) due to marine aerosols (see Fig. 7). The results based on observations from MODIS Aqua are similar. Over land, 50 % (0.117±0.093) of the total AOD$_{550}$ is anthropogenic, 35 % (0.090±0.102) is due to dust and 15 % (0.035±0.028) due to fine mode natural aerosols, while, over ocean, 40 % (0.079±0.080) of the total AOD$_{550}$ is of anthropogenic origin, 33 % (0.070±0.181) is due to dust and 27 % (0.054±0.018) due to marine aerosols (see Fig. 7). These results along with the relative contributions and the annual $\tau_a$, $\tau_d$, $\tau_n$ and $\tau_m$ levels for each one of the nine sub-regions of interest (see Fig. 1) are given in Table 4.

For anthropogenic aerosols, the region with the highest relative contribution is NBL (59 % for both Terra and Aqua MODIS) while the region with the lowest relative contribution is SWO (32 % for both Terra and Aqua MODIS) (see also Table 4). The spatial variability of $\tau_a$ is presented in Fig. 8a for Terra MODIS and Fig. S13a of the Supplement for Aqua MODIS, the patterns being similar in both cases. Over land, the annual $\tau_a$ patterns are similar to the AOD$_{550}$ patterns, the highest values appearing over local particle pollution sources (cities, industrial zones, etc.). Over the sea, $\tau_a$ is higher along the coasts, while it drops significantly towards other directions. An interesting feature here is that the oceanic region of Black Sea (BSO) presents higher relative anthropogenic contributions than the rest of the oceanic sub-regions but also than land areas with significant anthropogenic sources (e.g. ANL and NAL).

This is indicative of the transport of atmospheric particles from Central Europe and biomass burning aerosols during the biomass burning seasons in April-May from Russia (across the latitudinal zone 45°N-55°N) and July-August from South-Western Russia and Eastern Europe (Amiridis et al., 2010). These aerosols are transported at much lower latitudes as shown in previous studies (e.g. Vrekoussis et al., 2005; Karnieli et al., 2009) reaching the Sahara Desert and the Middle East regions (Pozzer et al., 2015). The fact that $\tau_a$ drops gradually from the coasts is also seen in Fig. 9 where the latitudinal variability of the optical depths of the different aerosol types ($\tau_a$, $\tau_d$, $\tau_n$ and $\tau_m$) is presented for four bands that cover the whole Eastern Mediterranean. An interesting feature is that $\tau_a$ increases nearby the shoreline (particularly along the North African coastal zone) before it gradually decreases. Over land aerosols are located within the atmospheric boundary layer, close to the emission sources, and hence, their deposition and removal from the atmosphere is more efficient than over the ocean, The particles which are transported over the ocean on the other hand usually reach greater heights which prolongs their lifetime.
As shown in Fig. 9, the same feature is observed for dust. Indicatively, \( \tau_d \) and the relative contribution of dust to the total AOD\(_{550} \) on an annual basis over the oceanic regions of SWO and SEO are in general higher or comparable to the ones over NAL (see Table 4 for more details). In Fig. 9, the MODIS-based \( \tau_d \) latitudinal variability is presented along with the latitudinal variability of dust AOD\(_{532} \) and extinction coefficients of dust at 532 nm from LIVAS. As expected, in all cases \( \tau_d \) decreases with distance from the large dust sources in the South and South-East (Sahara Desert, Middle East deserts) with local maxima over the latitudinal zone from 35°N to 40°N (especially for band 2 and band 3). The latitudinal variability of \( \tau_d \) is similar to the latitudinal variability of dust AOD\(_{532} \) for all the four bands despite the fact that the MODIS-based data have a resolution 100 times higher (0.1° vs 1°) and therefore are more sensitive to local characteristics. Dust reaches heights up to \( \sim 4-5 \) km in the area; however, the largest fraction of dust mass is confined within the first 2-3 km of the troposphere (see Fig. 9). The annual \( \tau_d \) patterns are shown in Fig. 8b for Terra MODIS (Fig. S13b of the Supplement for Aqua MODIS). The main dust transport pathways over the oceanic areas of the Eastern Mediterranean can be seen along with various local maxima over land. The highest \( \tau_d \) values over land appear over the regions of NAL and ANL (see Table 4) and along the coasts. The high dust concentrations appearing over these regions are not only due to the transport of dust from the nearby deserts but also due to local dust sources. A recent study by Liora et al. (2015) reports various local sources of wind blown dust along the coastal regions of Greece and Turkey, over the region of Anatolia in Turkey, over the Greek islands, Crete, Cyprus and regions close to the coastal zone of Middle East. Their results are in good agreement with the \( \tau_d \) patterns presented in this work.

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

Similar to fine mode natural aerosols over land, marine aerosols generally have the lowest contribution to the total AOD\(_{550} \) compared to the other aerosol types over the sea (see Fig. 7
and Table 4) except for BSO. The variability of $\tau_m$ is very low compared to $\tau_a$ and $\tau_d$. On an annual basis, high $\tau_m$ values appear over the Aegean Sea and the oceanic area between Crete and the North African coast while slightly lower values appear along the coasts of the Eastern Mediterranean (see Figs. 8d and 9). The $\tau_m$ patterns follow the near surface wind speed patterns in the region (see Fig. S14 of the Supplement) being in accordance to the $\tau_m$, marine particulate matter concentration or sea salt emission patterns appearing in other studies (Im et al., 2012; Nabat et al., 2013; Rea et al., 2015; Liora et al., 2015).

4.3.2 Seasonal contribution

The contribution of different aerosol types to the total AOD$_{550}$ over the Eastern Mediterranean varies from season to season. The relative contribution of each aerosol type over EML and EMO for each season is shown in Fig. 10. Over land, the relative contribution of $\tau_a$, $\tau_d$ and $\tau_n$ to the total AOD$_{550}$ exhibits a low seasonal variability. The relative contribution of anthropogenic aerosols to the total AOD$_{550}$ ranges from 49 % in SON to 50-55 % in DJF based on Terra MODIS observations and from 48 % in MAM and SON to 52 % in JJA based on Aqua MODIS observations. In contrast, over the oceanic regions the relative contribution of $\tau_a$, $\tau_d$ and $\tau_m$ to the total AOD$_{550}$ exhibits a significant seasonal variability. The relative contribution of anthropogenic aerosols to the total AOD$_{550}$ ranges from 27 % / 27 % in DJF to 50 % / 47 % in JJA based on Terra/Aqua MODIS observations. The percentages appearing here are in accordance to the values appearing in Hatzianastassiou et al. (2009) where a different satellite-based approach was followed. Indicatively, for the greater Athens area, an average summertime anthropogenic contribution of ~ 50 % was found here based on Terra MODIS data which is within the summer period range of 47-61 % indicated in the study by Hatzianastassiou et al. (2009). In addition, the corresponding values for the greater Thessaloniki area, Crete, Cairo and Alexandria are 53 %, 38 %, 48 % and 41 %, respectively, within the range of values (57-73 %, 36-52 %, 34-56 % and 23-60 %) shown in Hatzianastassiou et al. (2009). Only in the case of Ankara, our results suggest a lower anthropogenic contribution (52 % versus 71-84 %). Particularly for Athens, Gerasopoulos et al. (2011) following a different approach incorporating ground-based AOD observations and trajectory modeling reached similar results (annual contribution of ~ 62 % from local and regional sources and continental Europe which is expected to be mostly of anthropogenic origin). Similarly, for Crete, Bergamo et al. (2008) using a different approach, also utilizing ground-based data, found an annual anthropogenic contribution of ~ 43 %.
The seasonal patterns of the anthropogenic aerosols ($\tau_a$) over the Eastern Mediterranean based on MODIS Terra observations are presented in Figs. 11a, e, i and m while the seasonal variability of $\tau_a$ over the whole region, over the land covered part and the oceanic part and over the 9 sub-regions of interest is presented in Fig. 12. The results based on MODIS Aqua observations are similar and can be found in Figs. S15a, e, i and m and Fig. S16 of the Supplement. Generally, the local hot spots are detectable throughout the year; however, they are becoming much more discernible in spring and especially in summer. As shown in Fig. 12a, $\tau_a$ nearly doubles during the warm period of the year (spring-summer) with the seasonal variability being stronger over the sea (Fig. 12c) than over land (Fig. 12b). A clear peak is observed in summer, August being the month with highest $\tau_a$ levels. As discussed in Sect. 4.3.1 the summer peak is mostly a result of three basic reasons. The first one is the deficiency of wet removal processes compared to the cold period. As shown in Fig. S17, based on the TRMM satellite observations, August and July are the months with the lowest precipitation levels over the land covered part (a drop of ~ 75 % compared to winter months) and over the oceanic part (a drop of ~ 90 % compared to winter months) of the Eastern Mediterranean, respectively. The second reason is the enhancement of the photochemical production of secondary organic aerosols in summer (Kanakidou et al., 2011) and the third reason is the transport of pollution aerosols from Central Europe and biomass burning aerosols from South-Western Russia and Eastern Europe during the biomass burning season in July-August (Amiridis et al., 2010). The Etesians, which are persistent northerly winds that prevail over the Eastern Mediterranean during summer, bring dry and cool air masses and aerosols from the regions mentioned above while blocking at the same time the transport of desert dust in the region and dispersing local pollution in urban areas to levels typical for rural areas (see Tyrlis and Lelieveld, 2013 and references therein). As seen in Figs. 12a-l, a smaller but distinct in most cases $\tau_a$ peak appears in April mostly as a result of the transport of biomass burning aerosols from Russia (across the latitudinal zone 45°N-55°N). This is in line with the findings of Sciare et al. (2008) who detected traces of these biomass burning aerosols at the island of Crete in Southern Greece.

As discussed above, the relative contribution of dust to the total AOD$_{550}$ over land exhibits a low seasonal variability ranging from 29 % in DJF to 36 % in SON based on Terra MODIS observations and from 33 % in JJA to 38 % in SON based on Aqua MODIS observations (see Fig. 10). Over the oceanic regions the relative contribution of dust to the total AOD$_{550}$ ranges significantly throughout a year from 26 % / 28 % in JJA to 42 % / 39 % in MAM based on
Terra/Aqua MODIS observations. The percentages appearing here are in accordance to model and observational studies. For example, de Meij et al. (2012) using the atmospheric chemistry general circulation model EMAC (ECHAM/MESSy Atmospheric Chemistry) showed that dust contributes on an annual level ~ 30 % to the total AOD$_{550}$ over stations located in the area of the Eastern Mediterranean. Gerasopoulos et al. (2011) found a ~ 23 % percent contribution of North African dust to the total AOD over Athens using ground-based AOD observations and trajectory modeling. Taking into account that part of the ~ 39 % local and regional sources appearing in Gerasopoulos et al. (2011) is due to local dust sources, especially in summer, turns out that their results are in agreement with the ~ 33 % relative contribution found in this work for the greater Athens area based on Terra MODIS observations. The seasonal patterns of dust ($\tau_d$) over the region based on Terra MODIS observations are shown in Figs. 11b, f, j and n while the seasonal variability of $\tau_d$ over the whole region, over land, over the sea and over the 9 sub-regions of interest is shown in Fig. 12. The corresponding results based on MODIS Aqua observations are pretty similar and can be found in Figs. S15b, f, j and n and Fig. S16 of the Supplement. As seen in Fig. 11f, in spring, mostly due to the strong Sahara dust events, very high $\tau_d$ values appear over land regions in North Africa, Middle East, Anatolia and oceanic areas across the Eastern Mediterranean (especially below 35°N). Dust loading over the sea exhibits two maxima, one at the coastal zone of Libya and one across the coastal zone of Middle East. The same two maxima but with much lower $\tau_d$ values appear in summer (Fig. 11j) and autumn (Fig. 11n). Over land, the $\tau_d$ patterns are similar in summer and autumn, the maximum values appearing over the Anatolian Plateau and areas of North Africa and Middle East. During winter, dust maxima appear across the coastal zone of Northern Africa with relatively low $\tau_d$ values across the coastal zone of Middle East (Fig. 11b). In winter $\tau_d$ levels are low over land compared to the other seasons (Figs. 11b, f, j and n) as precipitation levels (see Fig. S17 of the Supplement) and hence wet scavenging of aerosols peak. At the same time, the local emissions of dust are low for regions away from the large area sources in the South (Liora et al., 2015). In contrast, over the sea $\tau_d$ levels in winter are similar or slightly higher for some areas than that in summer and autumn (see Figs. 11 and 12) as this is the season with the second highest frequency (after spring) of strong (~ 21 %) and extreme (~ 26 %) desert dust episodes in the region (see Gkikas et al., 2013 for details). February is by far the winter month with the highest $\tau_d$ levels (see Fig. 12) in line with the findings of Pey et al. (2013) who showed that the intensity of African dust episodes over stations in Greece and Cyprus peaks
Dust exhibits a strong peak in spring, April being the month with the highest \( \tau_d \) levels in line with other studies (e.g. Israelevic et al., 2012; Varga et al., 2014). The peak in April is a result of the high cyclonic activity over North Africa during this month as shown by Flaounas et al. (2015). According to the same study, low pressure systems are responsible for \( \sim 10-20\% \) of moderate and \( \sim 40-50\% \) of high and extreme Sahara dust transport events over the Eastern Mediterranean. North Africa (Sharav) cyclones develop mainly in spring and summer while Mediterranean cyclones develop in winter and autumn. The Mediterranean cyclones are more intense than Sharav cyclones. The region, is also affected by events bringing particles from dust source regions in the eastern part of the Mediterranean basin (Negev desert in Israel, Sinai in Egypt, Anatolian Plateau in Turkey) and the Arabian deserts (Basart et al. 2009; Pey et al., 2013; Abdelkader et al., 2015). Dust remains in the atmosphere for a period of 1-4 days undergoing chemical aging before being removed (see Abdelkader et al. 2015 and references therein). The seasonal variability of \( \tau_d \) is much stronger and the spring maxima much more prominent over the sea (see Fig. 12). This is expected, as dust is only occasionally transported over the sea during episodic events, while over land, local sources also contribute to the dust burden especially in summer due to the dryness of soil. For example, over NBL, a broad spring-summer peak is observed, June being the month with the highest \( \tau_d \) levels. As one moves south (SBL, ANL and NAL) the April peak becomes more prominent.

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

The seasonal relative contribution of marine aerosols to the total AOD\textsubscript{550} over the oceanic regions of the Eastern Mediterranean is shown in Fig. 10. \( \tau_m \) ranges from 20 \% in MAM to 35 \% in DJF based on Terra MODIS observations and from 21 \% in MAM to 36 \% in DJF based on Aqua MODIS observations (see Fig. 10). Like in the case of fine mode natural aerosols, the seasonal variability is very low, but here the highest values appear in winter (Fig. 12). Due
to the linear relation of $\tau_m$ and near surface wind speed within our algorithm (see Fig. 2) the
$\tau_m$ seasonal variability and patterns follow the wind speed ones (see Figs. 11d, h, l, p and
S14). Marine aerosol concentrations are lower close to the coastlines while the highest
concentrations (see Liora et al., 2015) and $\tau_m$ values within the Eastern Mediterranean appear
over the Aegean Sea (see Fig. 11). Overall, the $\tau_m$ patterns are in accordance to the $\tau_m$, marine
particulate matter concentration and sea salt emission patterns from previous studies (Im et
al., 2012; Nabat et al., 2013; Rea et al., 2015; Liora et al., 2015).

5 Summary and Conclusions
In this work, satellite data from MODIS Terra (3/2000-12/2012) and Aqua (7/2002-12/2012)
were analyzed separately in order to examine the spatial and temporal variability of aerosols
over the region of Eastern Mediterranean. A high resolution ($0.1^\circ \times 0.1^\circ$) MODIS gridded
dataset was compiled using a method that could be used in future regional studies. A number
of tests were implemented and the dataset was validated in detail using sunphotometric
observations from 13 AERONET stations. It is shown that the gridding method we use offers
the best compromise for studying the spatial variability of aerosols on a regional or local
scale, preserving at the same time the representativeness of the real aerosol load over each
specific spot.

Based on MODIS observations the average AOD$_{550}$ levels over the region of the Eastern
Mediterranean are $\sim 0.22 \pm 0.19$ which is $\sim 45$ % higher than the global mean. A number
of aerosol hot spots that coincide with megacities, large and even medium size cities, industrial
zones, power plant complexes, river basins, etc., can be detected on the AOD maps. A
number of local aerosol sources can also be seen on satellite retrieved tropospheric NO$_2$ and
planetary boundary layer SO$_2$ maps from OMI/AURA. This is indicative of the strong
presence of anthropogenic aerosols over these regions. Topography and precipitation also
play an important role. Generally, regions with mountain ranges are characterized by low
AODs while regions of low altitude are characterized by higher AODs. Regions with high
AOD$_{550}$ are in many cases connected to low precipitation levels and vice versa as
Precipitation is the major washout mechanism of atmospheric pollutants. Low removal rates
from precipitation contribute in preserving high the AOD$_{550}$ levels which are a result of
emissions and other atmospheric processes.

The AOD$_{550}$ patterns over the Eastern Mediterranean exhibit a significant seasonal variability
which is mostly driven by precipitation, photochemical production of secondary organic
aerosols, transport of pollution and biomass burning aerosols from Central and Eastern
Europe and transport of dust from the Sahara Desert and the Middle East. Differences
between MODIS Terra and Aqua Collection 051 AOD$_{550}$ over the Eastern Mediterranean are
generally small (~ -8 % over land and ~ 5 % over the sea). The comparison of the Terra-Aqua
differences with diurnal variabilities from the AERONET stations showed that only a part of
the observed differences is due to the diurnal variability of aerosols.
The MODIS data were combined with data from other satellites (Earth Probe TOMS,
OMI/AURA), reanalysis projects (ERA-Interim, MACC) and a chemistry-aerosol-transport
model (GOCART) to calculate the contribution of different types of aerosols to the total
AOD$_{550}$. The algorithm used was optimized for the Eastern Mediterranean through a number
of tests and comparison with LIVAS CALIOP/CALIPSO dust retrievals and AERONET
ground-based observations. A different approach is used for land and ocean covered areas
as there is not any reliable satellite retrieved quantity to separate the contribution of fine and
coarse mode aerosols over water surfaces.
Overall, for the land areas, based on MODIS Terra observations, 52 % (0.112±0.087) of the
total AOD$_{550}$ is due to anthropogenic aerosols, 32 % (0.074±0.080) due to dust and 16 %
(0.034±0.026) due to fine mode natural aerosols (see Fig. 7). For the oceanic areas, 41 %
(0.086±0.085) of the total AOD$_{550}$ is due to anthropogenic aerosols, 34 % (0.076±0.185) due
to dust and 25 % (0.054±0.018) due to marine aerosols. The results based on observations
from MODIS Aqua are in accord with previous studies.
Over land, the $\tau_a$ maxima are detected over local particle pollution sources (cities, industrial
zones, etc.). Over the sea, $\tau_a$ is higher along the coasts being significantly lower at greater
distance. Very high $\tau_d$ values appear over land regions in North Africa, Middle East, Anatolia
and oceanic areas across the Eastern Mediterranean, especially for latitudes below 35°N. Over
the sea, dust loading exhibits two maxima, one at the coastal zone of Libya and one across the
coastal zone of the Middle East. $\tau_d$ decreases with distance from the large dust sources in the
South and South-East. Generally, dust reaches heights up to ~ 4-5 km in the area, the largest
fraction of dust mass being confined within the first 2-3 km of the troposphere. The spatial
variability of $\tau_a$ and $\tau_m$ is very low compared to $\tau_a$ and $\tau_d$, following the total AOD$_{550}$ patterns
and the near surface wind speed patterns, respectively.
Over land, the relative contribution of anthropogenic aerosols, dust and fine mode natural
aerosols to the total AOD$_{550}$ exhibits a low seasonal variability, while over the sea the relative
contribution of anthropogenic aerosols, dust and marine aerosols shows a significant seasonal variability.

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

Overall, it is suggested that the AOD₅₅₀, τₐ, τ₉₄, τₐ₀ and τₘ high resolution gridded dataset which was compiled in this work could be used in a number of future atmospheric and biological studies focusing on the region of the Eastern Mediterranean (e.g. satellite and ground-based studies on aerosol-cloud-radiation interactions, experimental and field campaign studies on aerosols and clouds and research on the impact of aerosols on human health and nature). It is also acknowledged that a future update of the results presented here using more recent releases of MODIS aerosol data (e.g. Collection 6) and aerosol reanalysis datasets (e.g. NASA’s Modern-Era Retrospective Analysis For Research And Applications Aerosol Re-analysis) would be a useful contribution.

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Programme MarcoPolo (grant number 606953, theme SPA.2013.3.2-01) and from the European Union’s Horizon 2020 research and innovation programme under grant agreement no. 654109. The authors express their gratitude to the teams that developed the algorithms and produced the satellite products used in this study and to those who worked on the production of the model and reanalysis data used here. Special thanks are expressed to NASA Goddard Space Flight Center (GSFC) Level 1 and Atmosphere Archive and Distribution System (LAADS) (http://ladsweb.nascom.nasa.gov) for making available the MODIS Terra and Aqua Collection 051 level-2 aerosol data and the principal investigators and staff maintaining the 13 AERONET (http://aeronet.gsfc.nasa.gov) sites used in the present work. LIVAS has been financed under the ESA-ESTEC project LIVAS (contract no. 4000104106/11/NL/FF/fk). We thank the ICARE Data and Services Center (www.icare.univ-lille1.fr) for providing access to NASA’s CALIPSO data and acknowledge the use of NASA’s CALIPSO data. Special thanks are expressed to ECMWF (www.ecmwf.int) for the provision of the ERA-Interim and MACC reanalysis data. NASA’s GIOVANNI web database (http://giovanni.gsfc.nasa.gov/giovanni/) is highly acknowledged for the provision of Aerosol Index data from Earth Probe TOMS and OMI, aerosol data from the GOCART chemistry-aerosol-transport model (older version of GIOVANNI), tropospheric NO$_2$ and PBL SO$_2$ columnar data from OMI and precipitation data from 3B43 TRMM and Other Sources Monthly Rainfall Product. A.K.G. acknowledges the fruitful discussions with various colleagues from the Max Planck Institute for Chemistry and the Cyprus Institute (EEWRC) who indirectly contributed to this research.

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

<table>
<thead>
<tr>
<th>AERONET Station</th>
<th>Lat (°N)</th>
<th>Lon (°E)</th>
<th>Period of study</th>
<th>Country</th>
<th>Type</th>
<th>TERRA overpass</th>
<th>AQUA overpass</th>
</tr>
</thead>
<tbody>
<tr>
<td>Athens (NGA) (ATH)</td>
<td>37.988</td>
<td>23.775</td>
<td>05/2008-10/2012</td>
<td>Greece</td>
<td>Urban (coastal)</td>
<td>9:23±22min UT</td>
<td>9:32±22min UT</td>
</tr>
<tr>
<td>Bucharest (BUC)</td>
<td>44.548</td>
<td>26.030</td>
<td>07/2007-09/2012</td>
<td>Romania</td>
<td>Sub-urban (coastal)</td>
<td>9:17±24min UT</td>
<td>9:17±24min UT</td>
</tr>
<tr>
<td>CUT-TEPMAR (CUT)</td>
<td>34.675</td>
<td>33.043</td>
<td>04/2010-12/2012</td>
<td>Cyprus</td>
<td>Urban (coastal)</td>
<td>9:45±23min UT</td>
<td>10:31±23min UT</td>
</tr>
<tr>
<td>Elene (EFO)</td>
<td>44.075</td>
<td>26.632</td>
<td>09/2009-12/2012</td>
<td>Romania</td>
<td>Rural (coastal)</td>
<td>9:00±21min UT</td>
<td>10:04±22min UT</td>
</tr>
<tr>
<td>FORTH Crete (FOR)</td>
<td>35.333</td>
<td>25.282</td>
<td>01/2003-08/2011</td>
<td>Greece</td>
<td>Rural (coastal)</td>
<td>9:12±24min UT</td>
<td>10:15±25min UT</td>
</tr>
<tr>
<td>IMS-METU-ERDEMLI (IMS)</td>
<td>36.565</td>
<td>34.235</td>
<td>01/2004-01/2012</td>
<td>Turkey</td>
<td>Rural (coastal)</td>
<td>8:39±23min UT</td>
<td>10:49±22min UT</td>
</tr>
<tr>
<td>Lace University (LEC)</td>
<td>40.335</td>
<td>18.111</td>
<td>03/2003-12/2012</td>
<td>Italy</td>
<td>Sub-urban (coastal)</td>
<td>9:44±25min UT</td>
<td>11:49±25min UT</td>
</tr>
<tr>
<td>Nicosia (NSI)</td>
<td>31.922</td>
<td>34.789</td>
<td>02/2000-12/2012</td>
<td>Israel</td>
<td>Sub-urban (coastal)</td>
<td>8:30±24min UT</td>
<td>10:46±22min UT</td>
</tr>
<tr>
<td>Seydisehir (SED)</td>
<td>30.855</td>
<td>34.782</td>
<td>01/2006-04/2012</td>
<td>Israel</td>
<td>Rural (semi-arid)</td>
<td>8:30±27min UT</td>
<td>10:50±25min UT</td>
</tr>
<tr>
<td>Savusavu (SEV)</td>
<td>44.616</td>
<td>33.517</td>
<td>05/2006-12/2012</td>
<td>Uzbekistan</td>
<td>Urban (coastal)</td>
<td>8:51±21min UT</td>
<td>10:46±22min UT</td>
</tr>
<tr>
<td>Thessaloniki (THE)</td>
<td>40.650</td>
<td>22.960</td>
<td>09/2005-12/2012</td>
<td>Greece</td>
<td>Urban (coastal)</td>
<td>9:20±25min UT</td>
<td>11:33±22min UT</td>
</tr>
<tr>
<td>TUBITAK UZAY Ankara (TUB)</td>
<td>39.891</td>
<td>32.778</td>
<td>12/2009-04/2012</td>
<td>Turkey</td>
<td>Urban (continental)</td>
<td>8:48±26min UT</td>
<td>10:56±24min UT</td>
</tr>
<tr>
<td>Xanthi (XAN)</td>
<td>41.147</td>
<td>24.919</td>
<td>01/2008-10/2010</td>
<td>Greece</td>
<td>Rural (coastal)</td>
<td>9:01±25min UT</td>
<td>10:12±23min UT</td>
</tr>
</tbody>
</table>


Table 2. Results of the comparison of spatially (using a spatial window around each station) and temporally (±30 min from the MODIS overpass time) collocated MODIS Terra and Aqua (Italics) Collection 051 level-2 and AERONET sunphotometric (quadratically interpolated) AOD$_{550}$ observations for the Eastern Mediterranean stations. The algorithms used for the production of the validated MODIS data (DT and DB), the spatial window used for the spatial collocation (25 x 25 km$^2$ or 50 x 50 km$^2$ window around each station) with the AERONET data, the average MODIS and AERONET AOD$_{550}$ and the corresponding ±1σ values, the mean difference between them, the normalized mean bias (NMB) and the corresponding root mean squared (RMS) error, the percentage of the collocation points that fall within the expected error (EE) envelope and the pre-launch expected error (plEE) envelope (Expected Uncertainty - EU envelope for DB data), the correlation coefficient R, the slope a and the intercept of the regression line and the number of the collocation points are given in the table. L10 denotes the use of a collocation window of 50° x 50° as in Levy et al. (2010) while HQ denotes the use of high quality data only.
<table>
<thead>
<tr>
<th>Alg.</th>
<th>Window</th>
<th>MODIS TERRA AOD</th>
<th>AERONET</th>
<th>Mean Diff</th>
<th>NMB %</th>
<th>RMS err. in EE %</th>
<th>in pl EE %</th>
<th>R</th>
<th>x</th>
<th>h</th>
<th>Obs</th>
</tr>
</thead>
<tbody>
<tr>
<td>DFT 25 km</td>
<td>0.223±0.165</td>
<td>0.200±0.123</td>
<td>0.023±0.106</td>
<td>11.59</td>
<td>0.11</td>
<td>63.28</td>
<td>67.78</td>
<td>0.76</td>
<td>1.007</td>
<td>0.022</td>
<td>6697</td>
</tr>
<tr>
<td>DFT 25 km</td>
<td>0.247±0.173</td>
<td>0.197±0.121</td>
<td>0.050±0.109</td>
<td>25.18</td>
<td>0.12</td>
<td>57.14</td>
<td>62.87</td>
<td>0.78</td>
<td>1.113</td>
<td>0.027</td>
<td>6283</td>
</tr>
<tr>
<td>DFT 50 km</td>
<td>0.204±0.132</td>
<td>0.194±0.124</td>
<td>0.010±0.085</td>
<td>5.10</td>
<td>0.09</td>
<td>66.76</td>
<td>70.43</td>
<td>0.82</td>
<td>1.018</td>
<td>0.026</td>
<td>3357</td>
</tr>
<tr>
<td>DFT 50 km</td>
<td>0.224±0.133</td>
<td>0.194±0.123</td>
<td>0.030±0.090</td>
<td>13.34</td>
<td>0.09</td>
<td>46.67</td>
<td>50.43</td>
<td>0.82</td>
<td>1.018</td>
<td>0.026</td>
<td>3357</td>
</tr>
<tr>
<td>DB 25 km</td>
<td>0.220±0.177</td>
<td>0.186±0.128</td>
<td>0.040±0.162</td>
<td>21.38</td>
<td>0.17</td>
<td>-</td>
<td>51.90</td>
<td>0.47</td>
<td>0.657</td>
<td>0.104</td>
<td>2580</td>
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<tr>
<td>DB 25 km</td>
<td>0.242±0.237</td>
<td>0.182±0.118</td>
<td>0.060±0.196</td>
<td>33.03</td>
<td>0.20</td>
<td>-</td>
<td>55.30</td>
<td>0.44</td>
<td>0.815</td>
<td>0.094</td>
<td>3143</td>
</tr>
<tr>
<td>DBHQ 25 km</td>
<td>0.229±0.138</td>
<td>0.180±0.132</td>
<td>0.043±0.141</td>
<td>22.82</td>
<td>0.15</td>
<td>-</td>
<td>52.41</td>
<td>0.54</td>
<td>0.651</td>
<td>0.108</td>
<td>498</td>
</tr>
<tr>
<td>DBHQ 25 km</td>
<td>0.260±0.220</td>
<td>0.186±0.118</td>
<td>0.074±0.204</td>
<td>39.84</td>
<td>0.22</td>
<td>-</td>
<td>52.34</td>
<td>0.42</td>
<td>0.670</td>
<td>0.116</td>
<td>896</td>
</tr>
</tbody>
</table>

**Table 3.** AOD$_{550}$ levels, the corresponding ±1σ values and the number of gridded values used for the calculations over Eastern Mediterranean (EMT), over the land covered part (EML), over the oceanic part and over the 9 sub-regions of Eastern Mediterranean appearing in Fig. 1 based on the MODIS Terra and Aqua ( Italics) observations.
Table 4. Relative contribution of anthropogenic aerosols, dust, fine mode natural and marine aerosols to the total AOD$_{550}$ (bold) and the corresponding $\tau_a$, $\tau_d$, $\tau_n$, $\tau_m$ levels with their $\pm 1\sigma$ values (in parentheses) over Eastern Mediterranean (EMT), over the land covered part (EML), over the oceanic part and over the 9 sub-regions of Eastern Mediterranean appearing in Fig. 1 based on the MODIS Terra and Aqua (Italics) observations. The sum of the aerosol type AODs per region does not necessarily correspond to the total AOD$_{550}$ values appearing in Table 3 as these results were for the total of the days with aerosol retrievals even for days when our aerosol type separation algorithm was not applicable.
<table>
<thead>
<tr>
<th>Region</th>
<th>Satellite</th>
<th>Contribution to MODIS TERRA/AQUA MOD_\text{short}</th>
<th>Anthropogenic</th>
<th>Dust</th>
<th>Fine mode natural</th>
<th>Marine</th>
</tr>
</thead>
<tbody>
<tr>
<td>EML</td>
<td>TERRA</td>
<td>52 % (0.112±0.087)</td>
<td>32 % (0.074±0.080)</td>
<td>16 % (0.054±0.026)</td>
<td>-</td>
<td></td>
</tr>
<tr>
<td></td>
<td>AQUA</td>
<td>40 % (0.111±0.081)</td>
<td>22 % (0.062±0.083)</td>
<td>22 % (0.032±0.023)</td>
<td>-</td>
<td></td>
</tr>
<tr>
<td>EMO</td>
<td>TERRA</td>
<td>41 % (0.086±0.085)</td>
<td>34 % (0.076±0.181)</td>
<td>-</td>
<td>25 % (0.054±0.026)</td>
<td>-</td>
</tr>
<tr>
<td></td>
<td>AQUA</td>
<td>40 % (0.079±0.068)</td>
<td>33 % (0.070±0.181)</td>
<td>-</td>
<td>27 % (0.053±0.030)</td>
<td>-</td>
</tr>
<tr>
<td>NBL</td>
<td>TERRA</td>
<td>59 % (0.130±0.081)</td>
<td>23 % (0.042±0.080)</td>
<td>14 % (0.032±0.050)</td>
<td>-</td>
<td></td>
</tr>
<tr>
<td></td>
<td>AQUA</td>
<td>59 % (0.110±0.060)</td>
<td>24 % (0.035±0.067)</td>
<td>17 % (0.031±0.030)</td>
<td>-</td>
<td></td>
</tr>
<tr>
<td>SBL</td>
<td>TERRA</td>
<td>55 % (0.109±0.080)</td>
<td>29 % (0.065±0.080)</td>
<td>17 % (0.033±0.026)</td>
<td>-</td>
<td></td>
</tr>
<tr>
<td></td>
<td>AQUA</td>
<td>55 % (0.113±0.088)</td>
<td>29 % (0.060±0.060)</td>
<td>14 % (0.034±0.026)</td>
<td>-</td>
<td></td>
</tr>
<tr>
<td>ANL</td>
<td>TERRA</td>
<td>51 % (0.113±0.083)</td>
<td>34 % (0.076±0.064)</td>
<td>15 % (0.054±0.012)</td>
<td>-</td>
<td></td>
</tr>
<tr>
<td></td>
<td>AQUA</td>
<td>50 % (0.114±0.075)</td>
<td>34 % (0.079±0.070)</td>
<td>15 % (0.054±0.023)</td>
<td>-</td>
<td></td>
</tr>
<tr>
<td>NAL</td>
<td>TERRA</td>
<td>56 % (0.113±0.083)</td>
<td>35 % (0.083±0.085)</td>
<td>15 % (0.054±0.025)</td>
<td>-</td>
<td></td>
</tr>
<tr>
<td></td>
<td>AQUA</td>
<td>48 % (0.118±0.091)</td>
<td>38 % (0.099±0.100)</td>
<td>14 % (0.057±0.027)</td>
<td>-</td>
<td></td>
</tr>
<tr>
<td>BSO</td>
<td>TERRA</td>
<td>53 % (0.108±0.010)</td>
<td>22 % (0.064±0.010)</td>
<td>-</td>
<td>25 % (0.051±0.016)</td>
<td>-</td>
</tr>
<tr>
<td></td>
<td>AQUA</td>
<td>52 % (0.109±0.085)</td>
<td>22 % (0.062±0.083)</td>
<td>-</td>
<td>27 % (0.053±0.030)</td>
<td>-</td>
</tr>
<tr>
<td>NWO</td>
<td>TERRA</td>
<td>41 % (0.087±0.090)</td>
<td>33 % (0.071±0.142)</td>
<td>-</td>
<td>26 % (0.055±0.020)</td>
<td>-</td>
</tr>
<tr>
<td></td>
<td>AQUA</td>
<td>40 % (0.079±0.063)</td>
<td>32 % (0.060±0.127)</td>
<td>-</td>
<td>28 % (0.055±0.029)</td>
<td>-</td>
</tr>
<tr>
<td>SWO</td>
<td>TERRA</td>
<td>32 % (0.071±0.070)</td>
<td>42 % (0.097±0.237)</td>
<td>-</td>
<td>26 % (0.068±0.018)</td>
<td>-</td>
</tr>
<tr>
<td></td>
<td>AQUA</td>
<td>32 % (0.095±0.286)</td>
<td>41 % (0.072±0.186)</td>
<td>-</td>
<td>28 % (0.059±0.026)</td>
<td>-</td>
</tr>
<tr>
<td>NEO</td>
<td>TERRA</td>
<td>48 % (0.088±0.094)</td>
<td>28 % (0.061±0.144)</td>
<td>-</td>
<td>24 % (0.090±0.016)</td>
<td>-</td>
</tr>
<tr>
<td></td>
<td>AQUA</td>
<td>46 % (0.088±0.062)</td>
<td>28 % (0.057±0.113)</td>
<td>-</td>
<td>26 % (0.050±0.016)</td>
<td>-</td>
</tr>
<tr>
<td>SEO</td>
<td>TERRA</td>
<td>36 % (0.079±0.070)</td>
<td>39 % (0.087±0.124)</td>
<td>-</td>
<td>25 % (0.055±0.016)</td>
<td>-</td>
</tr>
<tr>
<td></td>
<td>AQUA</td>
<td>36 % (0.075±0.072)</td>
<td>38 % (0.080±0.217)</td>
<td>-</td>
<td>26 % (0.053±0.016)</td>
<td>-</td>
</tr>
</tbody>
</table>
Figure 1. Eastern Mediterranean map with the 9 sub-regions selected for the generalization of our results and the location of the AERONET stations used for the validation of MODIS satellite data. The 9 sub-regions are: NBL (Northern Balkans Land), SBL (Southern Balkans Land), ANL (Anatolia Land), NAL (Northern Africa Land), BSO (Black Sea Oceanic), NWO (North-Western Oceanic), SWO (South-Western Oceanic), NEO (North-Eastern Oceanic) and SEO (South-Eastern Oceanic). The full names and the geolocation of the 13 AERONET stations appearing in the map are available in Table 1.
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).
Figure 3. Comparison of spatially (using a 25 x 25 km$^2$ window around each station) and temporally (±30 min from the MODIS overpass time) collocated MODIS Collection 051 level-2 and AERONET sunphotometric (quadratically interpolated) AOD$_{550}$ observations for the Eastern Mediterranean stations: (a) for MODIS Terra DT data, (b) for MODIS Aqua DT data, (c) for MODIS Terra DB data and (d) for MODIS Aqua DB data. The color scale corresponds to the number of MODIS-AERONET collocation points that fall within 0.02 x 0.02 grid boxes. The solid line is the regression line of the MODIS-AERONET observations, the dashed-dotted line is the 1:1 line, the dotted lines represent the expected error (EE) envelope and the grey lines the pre-launch expected error (plEE) envelope (Expected Uncertainty - EU envelope for DB data). The slope and the intercept of the regression line, the correlation coefficient R, the normalized mean bias (NMB), the root mean squared (RMS) error, the percentage of the collocation points that fall within the EE and plEE and the number of all the collocation points.
Figure 4. AOD$_{550}$ patterns over Eastern Mediterranean as seen by MODIS Terra during the period 3/2000-12/2012 (3/2000-12/2007 for regions of North Africa covered by DB data only). The colorscale corresponds to the AOD$_{550}$ levels while the top x-axis and the right y-axis correspond to the longitude (°E) and latitude (°N), respectively. The position of 35 aerosol hot spots is marked on the map (numbers from 1 to 35) while the names of the places and the countries where the hot spots are located appear on the right of the map. In the same figure the exponential growth of the number of satellite-based articles focusing on aerosols over the greater Eastern Mediterranean from 1997 to 2014 is shown (black line). The black dots represent the number of articles published within three year intervals. The bottom x-axis and the left y-axis correspond to the years and the number of published articles, respectively. The exponential growth corresponds to a near doubling of the publication rate every 3 years.
Figure 5. (a) Tropospheric NO$_2$ levels and (b) Planetary boundary layer SO$_2$ levels (in $10^{15}$ molecules/cm$^2$) over the Eastern Mediterranean as seen from OMI/AURA (2005-2012), (c) Topography (GTOPO elevation data in meters above sea level) and (d) Annual precipitation levels (in mm/year) from 3B43 TRMM and Other Sources Monthly Rainfall Product (2000-2012).
Figure 6. Seasonal AOD$_{550}$ 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$_{550}$ on a seasonal basis appear on the right column.
Figure 7. Relative contribution of anthropogenic aerosols, dust and fine mode natural aerosols to the total AOD$_{550}$ over the land covered part of 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\textsubscript{550} over the oceanic part of Eastern Mediterranean based on MODIS Terra (b) and MODIS Aqua (d) observations.

Figure 8. (a) Anthropogenic aerosol (τ\textsubscript{a}), (b) dust (τ\textsubscript{d}), (c) fine mode natural aerosol (τ\textsubscript{n}) and (d) marine aerosol (τ\textsubscript{m}) patterns over the Eastern Mediterranean based on MODIS Terra observations during the period 3/2000-12/2012 (3/2000-12/2007 for regions of North Africa covered by DB data only).
Figure 9. Left column: Latitudinal variability of anthropogenic aerosols (τ_a), dust (τ_d), fine mode natural aerosols (τ_n) and marine aerosols (τ_m) for four 5-degree longitudinal bands (see embedded maps) covering Eastern Mediterranean based on MODIS Terra observations. Right column: Latitudinal variability of dust extinction coefficients at 532 nm in km⁻¹ (colorscale 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.
**Figure 10.** Seasonal relative contribution of anthropogenic aerosols, dust and fine mode natural aerosols to the total AOD$_{550}$ over the land covered part of 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$_{550}$ over the oceanic part of Eastern Mediterranean based on MODIS Terra (b, f, j, n) and MODIS Aqua (d, h, i, p) observations.

**Figure 11.** Seasonal (a, e, i, m) anthropogenic aerosol ($\tau_a$), (b, f, j, n) dust ($\tau_d$), (c, g, k, o) fine mode natural aerosol ($\tau_n$) and (d, h, i, p) marine aerosol ($\tau_m$) patterns over the Eastern Mediterranean based on MODIS Terra observations during the period 3/2000-12/2012 (3/2000-12/2007 for regions of North Africa covered by DB data only).
Figure 1. Seasonal variability of anthropogenic aerosols ($\tau_a$), dust ($\tau_d$), fine mode natural aerosols ($\tau_n$) and marine aerosols ($\tau_m$) over Eastern Mediterranean (EMT), over the land covered part (EML), over the oceanic part (EMO) and over the 9 sub-regions of the Eastern Mediterranean appearing in Fig. 1 based on MODIS Terra observations. The error bars represent the ±1σ values calculated from monthly gridded data.