Inventory of African desert dust events in the North-central Iberian Peninsula in 2003-2014 based on Sun photometer-AERONET and particulate mass-EMEP data

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Abstract. A reliable identification of Desert Dust (DD) episodes over North-central Spain is carried out based on AErosol RObotic NETwork (AERONET) columnar aerosol sun-photometer (aerosol optical depth, AOD, and Ångström exponent, α) and European Monitoring and Evaluation Programme (EMEP) surface particulate mass concentration (PMx, x=10, 2.5, and 2.5-10 µm) as main core data. The impact of DD on background aerosol conditions is detectable by means of aerosol load thresholds and complementary information provided by HYSPLIT (Hybrid Single Particle Lagrangian Integrated Trajectory Model) air mass back-trajectories, MODIS (Moderate Resolution Imaging Spectroradiometer) images, forecasting aerosol models, and synoptic maps, which have been carefully reviewed by a human observer for each day included in the DD inventory. This identification method allows the detection of low and moderate DD intrusions and also mixtures of mineral dust with other aerosol types by means of the analysis of α. During the period studied (2003-2014), a total of 152 DD episodes composed of 418 days are identified. Overall, this means ~13 episodes and ~35 days per year with DD intrusion, representing 9.65% days/year. During the identified DD outbreaks, 19 daily exceedances over 50 µg m⁻³ are reported at the surface. The occurrence of DD event days along during the year peaks in March and June with a marked minimum in April and lowest occurrence in winter. A large inter-annual variability is observed showing a statistically significant temporal decreasing trend of ~3 days/year. The DD impact on the aerosol climatology is addressed by evaluating the DD contribution in magnitude and percent (in brackets) for AOD, PM₁₀, PM₂.₅, and PM₂.₅-₁₀ obtaining mean values of 0.015 (11.5%), 1.3 µg m⁻³ (11.8%), 0.55 µg m⁻³ (8.5%) and 0.79 µg m⁻³ (16.1%), respectively. Almost similar annual cycles of DD contribution are obtained for AOD and PM₁₀ with present two maxima, one in summer (0.03 and 2.4 µg m⁻³ for AOD in June and PM₁₀ in August, respectively) and another in March (0.02 for AOD and 2.2 µg m⁻³ for PM₁₀), displaying both a similar evolution with exceptions in discrepancies occurring only in July and September. It is worth mentioning that the seasonal cycle of DD contribution to AOD does not follow the pattern of the total AOD (near bell shape), meanwhile both PM₁₀ cycles (total and DD contribution) present more similar shapes between them, although a main discrepancy is observed in September. The inter-annual

4

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The evolution of the DD contribution to AOD and PM$_{10}$ has evidenced a progressive decrease. This decline in the levels of natural mineral dust aerosols can explain up to the 30% of the total aerosol load decrease observed in the study area during the period 2003-2014. The relationship between columnar and surface DD contributions is evident with a correlation coefficient of 0.81 for the inter-annual averages. Finally, synoptic conditions during DD events are also analysed observing that the North-south African thermal low causes most of the events (~53%). The results presented in this study highlight the relevance of the area studied since it can be considered as representative of the clean background in the Western Mediterranean Basin where DD events have a high impact on aerosol load levels.

1 Introduction

Atmospheric aerosol particles play a key role in the radiation scattering and absorption physical processes that contribute to the Earth’s radiative budget (Trenberth et al., 2009; Wild et al., 2013). Their impact on Earth’s climate is represented by their direct radiative forcing (Haywood and Boucher, 2000; Boucher et al., 2013), but aerosols also act as cloud condensation nuclei (CCN)-modifying cloud properties and giving rise to a set of feedback processes that constitute the indirect radiative effect (Lohmann and Feichter, 2005; Lohmann et al., 2010; Boucher et al., 2013). All these aerosol climate effects have been enhanced due to anthropogenic aerosol particles (mainly sulphate and carbonaceous substances) which have increased the mean load of the global world in the last century and have modified substantially the atmospheric composition (Boucher et al., 2013). Aerosol radiative properties, such as aerosol optical depth (AOD) or single scattering albedo (d’Almeida et al., 1991; Cachorro et al., 2000; Eck et al., 2010) are important issues to consider when studying the impact of atmospheric aerosol on climate.

Beside the climatological aspect of atmospheric aerosols, another element to be considered is their direct incidence on air quality (Ganor et al., 2009; Kulmala et al., 2009; Ganor et al., 2009; Querol et al., 2013). In environmental studies, particulate matter is mostly represented by its level of mass concentration at the surface represented by various size fractions (PM$_{10}$, PM$_{2.5}$, PM$_{1}$, etc., where the subscript indicates the upper cut-off of the aerodynamic diameter of particles and PM$_x$ is used here as a general term referring to these fractions) and by the chemical speciation of its components—derived specific aerosol components: sulphates, nitrates, carbonaceous, mineral, among others. The different properties of aerosols are linked to derived effects like the aerosols can have a strong adverse impact on human health (e.g., Pope 2000; Pérez et al., 2012), and ecosystems (Mahowald et al., 2010).

Desert dust or mineral dust aerosol is one of the main natural types of atmospheric aerosol particles, with a strong impact on the Earth system due to its worldwide distribution and spatial-temporal variability (Goudie and Middleton, 2006; Knippertz and Stuut, 2014; Viana et al., 2014). The injections of Desert Dust (DD) into the atmosphere, from the main two Sahara’s major dust sources (Bodélé depression and Eastern Mauritania) by different re-suspension processes, can result in high layers being transported through very long distances to the Atlantic Ocean and Europe (e.g., Prospero et al., 1999; 2002; Escudero et al., 2006; Engelstaedter and Washington, 2007; 2011; Knippertz and Todd, 2012; Guirado et al., 2014).
Our interest in this work focuses on atmospheric aerosol studies over the Iberian Peninsula (IP), which constitutes a peculiar area due to the large spatio-temporal variability in aerosol properties, types and mixing processes as a result of the contrasting influences of the Atlantic Ocean, Mediterranean Sea, European continent and the Saharan area. Based on sun-photometer data studies, different sectors of the IP, basically defined by its topography/geography, can exhibit different aerosol climatologies (Alados-Arboledas et al., 2003; Vergaz et al., 2005; Estellés et al. 2007; Toledano et al., 2007a; Obregón et al., 2012; Bennouna et al., 2013; Mateos et al., 2014a). In particular, the Sahara and Sahel desert areas are the most important natural sources of mineral aerosols for the IP. The closeness of the IP to the African continent intensifies the impact of desert dust events on the aerosol load, measured on the whole atmospheric column (AOD) and at the surface (PMx). Different synoptic weather conditions and circulation patterns define-influence the arrival of desert dust intrusions in the IP with a different seasonal behaviour (Escudero et al., 2005, 2006; Toledano et al., 2007b; Basart et al., 2009; Valenzuela et al., 2012; Pey et al., 2013a; Salvador et al., 2013, 2014). These intrusions are characterized by isolated or episodic events of short duration (around 2-3 days). However, episodes in summer months are longer and more frequent in such a way that often successive episodes are linked due to the recirculation of air masses producing feedback processes, which give rise to a long residence time of desert dust particles in the atmosphere when there is low precipitation (Rodríguez et al., 2002; Escudero et al., 2005). Therefore, desert dust aerosols are one of the most important types over the IP, having an important influence on the air quality and radiative properties and hence detection, quantification and characterization are important research tasks. There are different ways to approach the detection or identification of desert dust events depending on the objectives of each study. The detection depends on the different techniques used: surface measurements, remote sensing (satellite or ground-based), back-trajectories evaluation, aerosol models, or a combination of them. Air mass trajectories have been one of the first and most used techniques to identify the origin of the transport of mineral dust aerosols to different regions worldwide (e.g., Hogan and Rosmond, 1991; Prospero et al., 2002; Kallos et al., 2003; Pace et al., 2006). Although there is abundant literature about mineral dust over Southern Europe or the Mediterranean areas, most of the studies about detection, characterization and/or impact of desert dust aerosols are focused on case studies or particularly strong episodes: e.g., in Italy (e.g., Meloni et al., 2007; di Sarra et al., 2011; Bègue et al., 2012), Greece (e.g., Kaskaoutis et al., 2008) or the Iberian Peninsula (e.g., Lyamani et al., 2005; Cachorro et al., 2006, 2008; Pérez et al., 2006; Cachorro et al., 2006, 2008; Guerrero-Rascado et al., 2009; Córdoba-Jabonero et al., 2011). Few studies are based on long-term datasets of desert dust using different techniques, such as sun photometers (Toledano et al., 2007b; Valenzuela et al., 2012), satellite sensors (Kaufman et al., 2005, Kaskaoutis et al., 2012; Gkikas et al., 2013; 2015), LIDAR measurements (Mona et al., 2006; 2014; Papayannis et al., 2008) or PMx data (Escudero et al., 2005; Pey et al., 2013a; Salvador et al., 2013, 2014; Pey et al., 2013a; Rodríguez et al., 2015). As can be seen from this list only recent studies contain long-term data sets, and only in some of them is the net contribution of DD evaluated. The PMx observations provided by different networks have constituted one of the most frequent tools for the establishment of DD inventories (e.g., Escudero et al., 2005; Pey et al., 2013a; Salvador et al., 2013, 2014) in order to evaluate their
contribution to PMx levels demanded by the EU directives. The EU 2008/50/EC directive (EC, 2008) on air quality establishes threshold values for the concentration of particles with aerodynamic diameter below 10 (PM$_{10}$) and 2.5 (PM$_{2.5}$) µm: an annual mean and 24 hour mean of respectively 40 µg m$^{-3}$ and 50 µg m$^{-3}$ for PM$_{10}$ and an annual PM$_{2.5}$ average of 25 µg m$^{-3}$. In this sense it is necessary to know the contribution of natural and anthropogenic components to the total aerosol load. Therefore, the contribution of mineral dust in South-Europe is important because of the link between PM$_{10}$ exceedances and DD outbreaks.

Once the identification of DD African aerosols is carried out, the next task is to quantify their contribution to the total aerosol load. The evaluation of impact or the contribution of DD episodes to PMx data is viable by means of a chemical speciation analysis (Rodríguez et al., 2001, 2002, 2015) but this method requires high man power and presents poor temporal sampling. Hence, in order to avoid this expensive technique other methods have been developed taking using PMx (Escudero et al., 2007; Ganor et al., 2009) and AOD data (Toledano et al., 2007b) or need to be developed. As reported by Viana et al., (2010) and taking into account more recent publications (e.g., MAGRAMA, 2013, 2015) no more than three methods are currently used with PMx including receptor models (Pey et al., 2013b; Belis et al., 2013). However, other techniques could such as receptor models widely applied for surface data (Pey et al., 2013b; Belis et al., 2013) may be updated or improved need to be updated to the measuring site for DD contribution estimates. In a similar way columnar aerosol algorithms can facilitate the apportioning of the different aerosol types (Dubovik et al., 2002; O’Neill et al., 2003) to the total aerosol load.

The advantage of remote sensing techniques, such as sun-photometry, for DD detection is the spectral information recorded by their AOD measurements and given by the Ångström exponent, α. This is a powerful tool in the identification and classification of the different aerosol types (Eck et al., 1999; Toledano et al., 2007a) but also allows “near-real-time” processing of data by means of reasonably sophisticated algorithms (Dubovik et al., 2002; O’Neill et al., 2003) to retrieve aerosol properties. The evolutions of a Surface PMx and columnar AOD quantities differ in the seasonal cycles present notable differences between them (see, e.g., Bennouna et al., 2014; Mateos et al., 2015) and hence their DD impact can also present some discrepancies in that cycle. The PMx sampling used here is based on daily filter records (see, Aas et al., 2013) while sun photometers provide instantaneous measurements of the columnar load but their sampling is limited to daytime sun-cloud-free conditions (Toledano et al., 2007a,b).

The usefulness of a DD inventory is that it opens the possibility of the evaluation of desert dust contribution to the total aerosol load. However, very few studies have accomplished this task over a long period and from a climatological multi-year perspective. To our knowledge, only the inventory of Toledano et al. (2007b) addressed the DD contribution to AOD between 2000 and 2005 in a Spanish South-western site (El Arenosillo). For PMx data we have found various studies, as those of more recent publication by Salvador et al. (2013; 2014) and Pey et al. (2013a). Salvador et al. (2013) reported a DD inventory and the corresponding contribution of DD is determined over the Madrid area from over the period 2001-2011, which this study is extended to several stations covering the whole IP by Salvador et al. (2014). Pey et al. (2013a), with the same methodology as Salvador et al. (2013), analysed the period 2001-2011 for PM$_{10}$ at different sites
over the whole Mediterranean Basin. In these three mentioned studies, the method followed for the evaluation of the DD contribution to PMx is different to that for AOD. Therefore, the development of methodologies for the evaluation of DD contribution is an open area of research, for instance, to obtain the near real-time DD contribution value.

Within this framework, the main purpose of this study is to establish a reliable inventory of DD episodes together with the evaluation of their contribution to the total aerosol load, using both AOD and PMx data. As an innovative aspect, this is the first time, to our knowledge, that DD events are identified by the simultaneous use of both columnar (AOD/α) and surface (PMx) aerosol observations. The methodology is applied over one of the cleanest atmospheric areas in South-western Europe, the North-central area ('Castilla y León' region) of the Iberian Peninsula, for a long time period spanning more than one decade (2003-2014). Regarding the columnar aerosol data, the reliable measurements performed within the Aerosol Robotic Network, AERONET (Holben et al., 1998), are used. For the study area, the only available aerosol long-term data set is recorded in the Palencia site (41.9° N, 4.5° W, and 750 m a.s.l.). Regarding surface aerosols, we use the high quality particulate matter data recorded by the European Monitoring and Evaluation Programme (EMEP) network in the nearby Peñausende station (41.28°N, 5.87°W, and 985 m a.s.l.), with similar background conditions to Palencia. In this way, the usage of these two worldwide extended and high-quality networks will ensure the feasibility of implementing the proposed method in other regions. The long-term inventory described hereafter has been employed to establish the main characteristics of the DD episodes in North-central Spain: their climatology, inter-annual behaviour, trends for the number of episodes and associated days, and occurrence under different synoptic scenarios. In addition, the evaluation of the DD contribution to the total mean values of AOD and PMx is also addressed over the period investigated from a climatological and inter-annual perspective, which emphasizes the correlations between both quantities.

Section 2 describes the region of study and the datasets used. The methodology followed in the DD identification and in the evaluation of its contribution to aerosol load is presented in section 3. In subsections 4.1-4.3, the seasonal cycles and inter-annual evolution of DD events, dusty days, and DD contribution to AOD and PMx are deeply investigated. Subsection 4.4 provides an estimation of the uncertainty of this method and subsection 4.5 describes the synoptic scenarios associated with the arrival of DD episodes to the North-central Iberian Peninsula. Finally, Section 5 summarizes the main findings obtained in this study.

2 Sites of measurements and database

2.1 AERONET network and AOD/Ångström exponent database

The main database for this study contains the instantaneous values of AOD obtained for 440 nm (henceforth AOD for shake of clarity) and Ångström exponent (α for the 440-870 nm range) measured in the Palencia site of the AERONET-Europe network (see Figure 1 and Table 1). The instrument used to obtain these data is a CIMEL CE318 radiometer which
measures under clear sky conditions and each every 15 minutes. The raw measurements in all sky conditions (level 1.0 of AERONET criteria), the cloud screened data (level 1.5), and the high quality processed data in the level 2.0 are used in this study. Lower quality levels are of great help to reliably determine the duration of each DD episode, in particular, when the event is mixed with cloudiness.

Aerosol measurements are available in the Palencia site since 2003, one of the longest series of aerosol optical measurements in the Iberian Peninsula (Mateos et al., 2014a). The number of available days for every year in this study can be seen in Table 1. The standardization protocols of AERONET demand require pre- and post- calibration of the instruments after a field measuring period of 12 months, which helps to assure the quality of the obtained data and associates an uncertainty for AOD_{440nm} of ±0.01 and for the derived α parameter of about ±0.03 (see, e.g., Toledano et al., 2007a).

The Palencia site is placed in the autonomous region of “Castilla y León” in the North-central Iberian Peninsula, which is also known as the “Castilian plateau” with an average altitude of ~800 m. This region is the third less-populated community in Spain due to its big large area (94,193 km²) and its low population (2,543,413 inhabitants registered in the census in 2012), with a population density just a bit higher than 27 inhabitants per km². Palencia is a small city (100,000 inhabitants) placed located in the north of “Castilla y León” but the measuring site is located outskirts being surrounded by rural areas, removed from big urban and industrial centres. Hence, this area exhibits an exceptionally clean atmosphere and aerosol observations are representative of the background conditions for the whole region. Therefore, desert dust intrusions are can be observed since they notably significantly modify the background aerosol properties.

In order to fill the gaps in the Palencia AOD database, the site “Autilla” (42.00ºN, 4.60ºW, and 873 m a.s.l.) close to Palencia (7 km apart) has been used (see for details see Bennouna et al., 2013). This site is being used by GOA (Grupo de Optica Atmosférica) as the calibration platform for Cimel sun photometers within the AERONET-EUROPE infrastructure and also works as a routine measurement site. Under these considerations, the columnar aerosol data series used in this study is consistent and allows one to perform the inventory of DD events in this region.

2.2 EMEP network and PM database

Daily PM_{10} and PM_{2.5} measurements provided by the EMEP network constitute the second core database used to carry out this study (see Figure 1 and Table 1). This network has the objective of regularly providing qualified scientific data to the interested organizations in order to analyse and assess the transboundary transport and emission of pollutants (e.g., Aas et al., 2013). This is the objective of the LRTAP (Long-Range Transboundary Air Pollution) convention that establishes a framework for cooperative action for reducing the impact of air pollution. Using the PM_{10} and PM_{2.5} data, the particulate matter associated with coarse particles (PM_{2.5-10}) can be determined by subtracting these quantities (PM_{10} – PM_{2.5}). PM_{10} and PM_{2.5} data belonging to the EMEP site of Peñausende have also been used for the detection of dusty days with DD load with AOD–α data of the Palencia site, and for the evaluation of DD their contribution to the total PM_{x} levels. However, in order to better detect the DD episodes that arrive to our study area, two nearby stations (Campisábalos, 41.28ºN, 3.14ºW,
and 1360 m a.s.l., and Barcarrota, 38.48°N, 6.92°W, and 393 m a.s.l., see Figure 1) are taken as complementary sites. All these three sites (Peñausende, Campisábalos, and Barcarrota) are placed in rural areas where background values are measured and the detection of Saharan desert dust intrusions is also possible. Among all Spanish EMEP sites, these two complementary sites have not been randomly selected, since their geographical locations help us to have more information about the path followed by the intrusion before arriving in our study area. The arrival direction of arrival can be established by measurements from the west (by the Peñausende site), South-southwest (by the Barcarrota site) and South-southeast (by the Campisábalos site).

The available time period for the PMx data starts in 2001, but the same period (2003-2014) used in the AOD$_{440nm}$ is considered for the homogenization of the results presented in this study. It is important to emphasize here that in spite of the distance between the Peñausende and Palencia sites (~100 km), the absence of any large landforms between them together with their atmospheric and background conditions make possible the joint discrimination and evaluation of these observations of AOD and PMx for the detection of DD intrusions. Furthermore, a deep analysis about the air masses at Peñausende and Palencia sites (not shown here) has been carried out, corroborating that the geographical distance between them is negligible in for the analysis of regional quantities such as AOD, water vapour, and ozone column, among others. The use of AOD$_{440nm}$, PM$_{10}$, and PM$_{2.5}$ observations provides a complete and detailed comprehensive database to carry out an analysis in terms of aerosol load and particle size, both at the surface and in the whole atmospheric column. Table 1 presents a detailed description of the number of days with available data every year of each database annual sampling of each database. Overall, the PMx temporal sampling represents a larger amount of days with data than for AOD. Particularly, the year 2003 presents the lowest AOD sampling (42% of 365 days) because of the certain gaps at the start of the sun photometer measurements.

### 2.3 Ancillary information

With the aim of carrying out a more accurate evaluation and discrimination of days that constitute a desert dust intrusion, ancillary information is taken into account also considered. Air mass back-trajectories arriving at Palencia at 12.00 UTC have been calculated with the HYSPLIT model (Hybrid Single-Particle Lagrangian Integrated Trajectory), version 4 (Draxler et al., 2014; Stein et al., 2015). Due to the fact that desert dust aerosols can be transported to altitude levels higher than the boundary layer, back-trajectories have been calculated for three heights (500, 1500 and 3000 meters above ground level) and analysed 5 days back in time (120 h), assuming using the model vertical velocity in the calculations. The meteorological database used as input for HYSPLIT is the Global Data Assimilation System (GDAS) GDAS1 dataset (e.g., Su et al., 2015). These three levels are commonly used in these studies to represent levels are very usual in this type of studies to represent the air masses near the surface, in the boundary layer and in the free troposphere, in order to follow the vertical transport of aerosols. Important information about cloudiness is obtained from MODIS (Moderate Resolution Imaging Radiometer) rapid response imagery products (https://earthdata.nasa.gov/data/near-real-time-data/rapid-response). In addition, GIOVANNI (Geospatial...
Interactive Online Visualization AND aNalysis Infrastructure) MODIS AOD aerosol maps (http://giovanni.gsfc.nasa.gov/giovanni/) and those provided by the AERONET website (http://aeronet.gsfc.nasa.gov/cgi-bin/bamgomas_interactive) are used to determine the extension and path followed by the mineral dust air masses for a DD identified event. The NAAPS Global Aerosol model (Navy Aerosol Analysis and Prediction System; available at http://www.nrlmry.navy.mil/aerosol/) is also used to corroborate if model forecasts also detect a given DD episode over the study area.

As it has been analysed in previous studies (Escudero et al. 2005; Toledano et al. 2007b) desert dust intrusions over Spain take place under certain synoptic scenarios (see Section 4.5 for further details). Through the Earth System Research Laboratory of NOAA (National Oceanic and Atmospheric Administration), the plots of the geopotential height at 700 hPa and mean sea level pressure are obtain in order to evaluate the synoptic scenario associated for a given DD episode. For evaluating these scenarios, the geopotential height at 700 hPa and mean sea level pressure are required. Through the Earth System Research Laboratory of NOAA (National Oceanic and Atmospheric Administration), the plots of atmospheric quantities can be obtained for the selected days and one possible scenario among the four possibilities (see Section 4.5) proposed by Escudero et al. (2005) is identified.

3 Methodology

3.1 Detection of desert dust episodes

This study is based on instantaneous AOD$_{440\text{nm}}$ and $\alpha$ values, as well as daily PM$_{10}$ and PM$_{10}$/PM$_{2.5}$ ratio data. The method for the detection of Desert Dust (DD) intrusions is a manual inspection of the evolution of these four quantities together with the origin of the air masses at the three levels of altitude at 500, 1500, and 3000 m a.s.l. and the auxiliary material of AOD MODIS maps, aerosol models and synoptic scenarios. The methodology for detection is similar to that applied in Toledano et al., (2007b) with the added information of PMx data, and not so much different from that used in other studies also based on a set of different observations (Escudero et al., 2005; Pace et al., 2006; Kalivitis et al., 2007; MAGRAMA, 2013). The difference between these methods lies on the weight played by each quantity, or and the way to analyse the information. For example in our case the AOD-PMx data is the first and primary information, but in other studies the key variable is the origin of the air masses (Pace et al., 2006; Valenzuela et al., 2012). Meteorological products and forecast aerosol models can also be used for this task (MAGRAMA, 2013). Although automatic methods can be applied in the DD identification, a visual inspection should be performed to corroborate each classification.

This study has been carried out as a year by year service to the “Consejería de Medio Ambiente” of the Autonomous Community of ‘Castilla y León’ by means of two Research Programmes from 2006-2013 about the “Discrimination, characterization and evaluation of desert dust outbreaks over ‘Castilla y León’ region”. These programmes have the purpose of accomplishment programmes aim to help accomplish implement the “Environmental Quality Improvement Policy” of
the EU by the National and Regional Governments of Spain. The experience gained with this year by year identification gives rise to provides the final DD inventory presented in this study.

Certain thresholds have been established in order to identify those conditions which are separated from the background over the studied area. Hence, these choices are based on the aerosol climatology of the site. For Palencia, mean $\text{AOD}_{440nm}$ is 0.13 with a standard deviation (std) of 0.09 and $\alpha=1.27 \pm 0.36$. Consequently, DD intrusions are firstly detected using the following thresholds: $\text{AOD}_{440nm} \geq 0.2$ (AOD Criterion in Table 1), which is the mean plus the standard deviation approximately, and $\alpha \leq 1.0$ for instantaneous values. If a day has about 50% of these data, the day is considered as a dusty day and these events are called “pure desert aerosols”, denoted by (D). However, days with instantaneous values of $\text{AOD}_{440nm} \geq 0.2$ and $1.0 \leq \alpha \leq 1.5$ may also present desert dust aerosols but mixed with other types (indicated by the high $\alpha$ value), which is corroborated by the desert origin of air mass backwards trajectories, satellite images, and model forecasts. This type of days, named MD (Mixed Desert), may either be a part of an intense event (generally of D type) or form by themselves a low-moderate intensity event. Previous studies have also stated that DD intrusions in the Mediterranean Basin can present moderate AOD associated with large $\alpha$ values (e.g., Pace et al., 2006; Tafuro et al., 2006).

The limit of $\alpha < 1$ for identifying coarse particles has been established by previous studies in different areas (e.g., Eck et al., 1999, 2010; Dubovik et al., 2002; Meloni et al., 2007), making this threshold suitable for our study area. It is important here to emphasize that the $\alpha$ parameter allows a more fine identification of desert dust events, mainly those of low intensity with a less desert character (the larger the AOD of a desert event, the lower $\alpha$), generally those mixed with other type of aerosols which are not accounted for in many DD studies (e.g., Gkikas et al., 2013). DD events of low intensity are more difficult to detect and hence it is more difficult to evaluate their contribution or impact on the AOD and PMx daily values.

The criterion for the PM$_{10}$ values is selected from our investigations (Bennouna et al., 2014; Mateos et al., 2015) and previous results (see, e.g., Querol et al., 2009, 2014), giving a threshold of PM$_{10} = 13 \ \mu g \ m^{-3}$ for the daily values of the particulate mass concentration (PM Criterion in Table 1). The mean value of PM$_{10}$ over the period 2003-2014 is $10 \pm 9 \ \mu g \ m^{-3}$, thus 13 is the mean plus one third of the standard deviation. However, this value alone must be taken as a “warning value” and not as a true threshold, in the sense that this value alone does not define a dusty day. This PM$_{10}$ value could seem very low but we must note here that all these values are manually supervised by the expert observer who will take the final decision of the inclusion or not of a dusty day, having in mind all the available information given by these data and all the complementary material.

Certain thresholds have been established to identify those conditions which stand above the clean background over the study area. Hence, the choice of these thresholds is based on the aerosol climatology of the site from our investigations (Bennouna et al., 2014; Mateos et al., 2015) and previous results (see, e.g., Querol et al., 2009, 2014). The mean values for the long-term period 2003-2014 are $0.13 \pm 0.09$ for $\text{AOD}_{440nm}$ and $10 \pm 9 \ \mu g \ m^{-3}$ for PM$_{10}$. Hence, to detect the DD intrusions a visual inspection of the entire database is performed. When a day shows a group of number of points of the instantaneous AOD $\geq 0.18$ and/or the daily PM$_{10} \geq 13 \ \mu g \ m^{-3}$, that day is further investigated. The AOD threshold corresponds to the mean value plus half of the standard deviation, approximately, meanwhile that for PM$_{10}$ is the mean plus one third of the standard
deviation. Hence, these thresholds must be taken as “warning flags” in the sense that these values alone do not define the classification as a dusty day. They also need the ancillary information given by the air mass backwards trajectories, satellite images, weather maps, and model forecasts to determine and corroborate the origin of aerosols and synoptic conditions. Therefore, with all this information the human observer decides if a day must be included or not in the DD inventory.

In parallel to the above analysis, the evolution of the $\alpha$ quantity is also checked, allowing the identification of two different types of DD intrusions. Those days displaying $\alpha < 1.0$ in most of the instantaneous columnar data are identified as the “purser” desert dust intrusions and they are denoted by a D flag. Those days with $\alpha$ values in the interval $1.0 \leq \alpha \leq 1.5$ (which have been classified as dusty days by their aerosol load and/or the ancillary information) present a mixture with other types (e.g., clean continental aerosols) and they are denoted by a MD (Mixed Desert) flag. The MD event days may either be a part of an intense event (generally of D type) or form by themselves a low-moderate intensity event. Previous studies have also shown that DD intrusions in the Mediterranean Basin can present moderate AOD associated with large $\alpha$ values (e.g., Pace et al., 2006; Tafuro et al., 2006; Boselli et al., 2012). The limit of $\alpha \leq 1$ for identifying coarse particles has been established by previous studies in different worldwide areas (e.g., Eck et al., 1999; 2010; Dubovik et al., 2002, Meloni et al., 2007; Boselli et al., 2012), making this threshold suitable for our study area. It is important to emphasize here that the $\alpha$ parameter allows a more accurate identification of desert dust events, mainly those of low intensity, having less of the characteristics identifying a desert origin (overall, the larger the AOD of a desert event, the lower the $\alpha$), generally mixed with other aerosol types which are not accounted for in many DD studies (e.g., Gkikas et al., 2013). These DD events of low intensity are more difficult to detect because of the low signal of mineral dust aerosols and hence it is more difficult to evaluate their contribution or impact on AOD and PMx daily values. In fact, although the aerosol load threshold used in this study might seem very low, we must note again here that all these quantities are manually supervised by the expert-observer who will take the final decision of the inclusion or not of a dusty day, bearing in mind all the available information given by these data and all the complementary material.

Another problem to be considered in the identification of a dusty day is when AOD and PM$_{10}$ measurements show different information. For instance, AOD fingerprints seem to indicate possible desert dust presence meanwhile PM$_{10}$ not, e.g., AOD indicates desert dust presence but PM$_{10}$ does not. It must be taken into account that PM$_{10}$ quantity does not necessarily follow the same temporal behaviour as the AOD and possible time delays in PM$_{10}$ concentration due to deposition processes can occur. Desert dust events can reach the IP at high altitude layers (e.g., above 2000 meters, Gkikas et al., 2015), and dry deposition can last various hours/days. Assuming an average speed of deposition of around 0.6 cm s$^{-1}$ (Zender et al., 2003), DD particles can remain in the troposphere up to two days after an episode ends (Escudero et al., 2007). Hence, these possible day delays between columnar air masses and surface PM$_{10}$ are very variable (e.g., Kalivitis et al., 2007; Pey et al., 2013a). Therefore, as mentioned AOD and PM$_{10}$ observations must be considered as complementary information to detect mineral dust aerosols since with the evolution of both quantities it is easier to identify DD fingerprints.
It is worth mentioning here that in the general detection of both intensity and duration of a particular event, DD causes an increase in the AOD$_{440\text{nm}}$ and PM$_{10}$ values, which then surpass or just reach the corresponding threshold, together with a decrease in the $\alpha$ and PM$_{2.5}$/PM$_{10}$ values, giving rise to an increase of the mean size of the particle distribution. The duration of each intrusion can be established, since the background typical values, characteristics of the regional background, are recovered when the event finishes. The central or most intense days of each event are easy to detect due to the large increment of aerosol load with large particles ($\alpha$ even close to zero) but low to moderate events are more difficult to detect. Although the events vary in their nature, the first and last days of a DD event show a low or moderate signature of mineral dust particles because of its mixture with other aerosol types (e.g., clean continental aerosol in our study area), with the exception of the strong DD events which generally have a notable impact on the aerosol load levels since from the first day of the episode.

In the inspection of the instantaneous columnar dataset, non-reliable records are identified and removed due to their high dispersion in the data, likely attributed to cloud contaminated cloudy conditions. In the corroboration of some critical decisions made by the expert-observer the ancillary information (see section 2.3) constitutes a key point of the methodology. For instance, the verification of cloudy conditions can be supported by means of a comparison between AOD instantaneous data from different levels, 1.0 (all-sky conditions) and 1.5 (cloud-screened data), and the visualization of cloud systems in MODIS true colour and cloud product images. If there are signs of cloud presence, instantaneous AOD data are carefully checked to discern between non-valid cloud-contaminated data and a DD intrusion.

Once aerosol load measurements for a certain given day indicate the likely classification as a dusty day, the air mass back-trajectories (calculated as described in Section 2.3) are visualized in order to check if the origin of the path or the path followed crosses the North African region and/or its surroundings. Therefore, the air mass back-trajectory analysis and the geopotential maps (establishing a particular synoptic scenario, see Sections 2.3 and 4.5) lead to the final decision with respect to a DD event day classification, even in those days showing cloudiness. Finally, to help in the understanding of the general situation about the geographical distribution of the aerosol plumes, AOD MODIS maps and NAAPS forecasts are also inspected visualized for ensuring the final choice. The consistency of the information used provides a reliable identification of the DD event.

It is worth mentioning here that the final decision to include a day as D or MD is made by the human-observer with all the available information at hand. Perhaps this methodology is not the most adequate to apply for a big area with a high number of stations and long-term databases, but it is necessary for developing methodologies, because it will allow validation of other more automatic methods (e.g., those only based on threshold criteria), most of them using satellite observations (e.g., Gkikas et al., 2013, 2015).
3.2 Evaluation of desert dust contribution to total AOD and total PM$_{10}$ concentration

Once the DD inventory is established, the evaluation of the DD contribution can be addressed at seasonal and annual scales. Following Toledano et al., (2007b), the contribution of the DD events to AOD can be obtained as the difference of the multi-annual monthly means considering all days and the corresponding value without including the desert dust cases. This procedure was also used for PM$_{10}$ data in a 3-year evaluation of net DD contribution in several sites in the Mediterranean Basin (Querol et al., 2009). In this study, the annual cycle of the DD contribution to AOD/PMx is evaluated with the same methodology over the entire period 2003-2014, using the DD event days classified in the inventory. Furthermore, the relative DD contribution to AOD/PMx can be obtained by normalizing each one with respect to the corresponding total AOD/PMx value. Regarding the seasonal evaluations, the classification is used as follows: winter (Dec-Jan-Feb), spring (Mar-Apr-May), summer (Jun-Jul-Aug), and autumn (Sep-Oct-Nov). Analogously, the yearly AOD/PMx means excluding dusty days are subtracted from the yearly AOD/PMx means for all days for obtaining the DD contribution at annual time scale.

This method assumes the entire daily aerosol load (both surface and columnar) due to DD aerosols, being included the contribution of regional background aerosols also included, Thus, suitable time scales for this kind of DD contribution calculation are the annual and climatological multi-annual monthly means over 12 years of data. Those evaluations for every single day or month can be addressed using other methods, for instance the determination of percentile 40 of the time series without dusty days -to evaluate the background conditions that are subtracted from PMx levels (Escudero et al., 2007). This method has been taken as the standard by the European Commission for the evaluation of DD contribution to aerosol load at the surface (Viana et al., 2014; MAGRAMA, 2013, 2015).

The methodology used in this study leads to lower uncertainty in the annual cycle evaluation since there is good data coverage for the multi-annual monthly sampling (12 years). However, in the evaluation of year-by-year DD contribution to aerosol load, a given year can present a low coverage leading to a but higher uncertainty for a yearly evaluation. The inclusion or not of an uncertain DD event (e.g., very data contaminated with clouds, where cloud optical depth is assigned to aerosol AOD) can substantially modify the corresponding yearly mean as it has been shown in Bennouna et al. (2014). This source of uncertainty must be considered in the temporal trend evaluation. It is not easy to establish an adequate methodology to evaluate the DD aerosol contribution to aerosol load (Viana et al., 2010) and much less with its corresponding associated error. A further investigation is necessary about this subject. A discussion about the uncertainty of our approaches in the DD identification and in the evaluation of DD contribution to aerosol load can be found at the end of the results in subsection 4.4.
4 Results and discussion

4.1 Evaluation of the number of episodes and days: annual cycle and year-to-year variability

4.1.1 Evaluation of the number of episodes and dusty days

The inventory of desert dust intrusions includes: information on each episode and its associated days; the daily mean AOD, \( \alpha \), PM\(_{10} \), and PM\(_{2.5} \); cloudiness, synoptic scenarios, and air mass origin at the three altitude levels mentioned above (500, 1500, and 3000 m a.s.l.). Tables 1 and 2 show the information used to classify DD events and the main statistics for this inventory, respectively.

The PM\(_{10} \) sampling presents the best coverage of the measuring time period with a 93.1% of the days, AOD is available 67.2% of the time, and the coincident sampling is available 63.2% of the time. As can be deduced from Table 1, the majority (51.2%) of the DD event days composing the inventory are noticeable in both AOD and PM\(_{10} \) datasets. However, 46.3% of the total detected days are over the required thresholds only in one quantity (AOD or PM\(_{10} \)). This is the great advantage of the proposed inventory because DD outbreaks are identified with two complementary quantities about the aerosol load. The reasons behind this 46.3% (19.1% only with AOD and 27.2% only with PM\(_{10} \)) are due to time delays between columnar and surface levels related to deposition phenomena and also are due to the lack of AOD or PM\(_{10} \) measurements. Therefore, if the inventory is addressed by only one quantity, a large number of dusty days can be lost in their identification. Finally, a smaller number of cases (2.4%) are identified as dusty days using the ancillary information when AOD and PM\(_{10} \) data are not available.

The smaller amount of available data in the coverage of AOD time series (see Table 1), in comparison with PM\(_{10} \), is not a major handicap in DD detection. There are several years (2004, 2007, 2008, 2010, 2011, and 2012) with more DD days detected only by AOD than only by PM\(_{10} \), in spite of the smaller AOD sampling (between 53 and 103 days less per year). However, years 2003 and 2006, which present less than 200 daily AOD data, require the use of PM\(_{10} \) in order to better identify DD intrusions.

As reported in Table 2, during 2003-2014, a total number of 152 episodes have been identified, composed of 4189 days. Among them, 244242 days have been classified as days with desert aerosols (D) and 176 days with mixed aerosols (MD).

Overall, this means 13 episodes and 35 days per year with desert dust intrusion, representing the 9.65% of the days each year. The duration of DD episodes is very variable, ranging from 1 to 13 days, but a value of 2.7 days is obtained as the mean episode duration. Due to the high variability in the intensity of these intrusions it is difficult to distinguish when an event has ended or its intensity has simply fallen below our threshold. During summer, the recirculation of air masses in the IP is very frequent and the DD episodes are subject to large intensity variations. We have considered separate DD episodes when there is, at least, one day that does not meet the DD requirements between two DD episodes.

Our percentage of dusty days of 9.65% is lower than that reported by Salvador et al. (2013), which is around 18% (18 episodes and 65 days per year), who analysed DD intrusions over the central Iberian Peninsula (Madrid area) between 2001 and 2008. This large difference between two nearby areas (separated by ~200 km) can be explained by the different time
periods considered and the existence in between these areas of a high mountain range (Sistema Central), with peaks up to 2400 m (a.s.l.). For the North-eastern area of the Iberian Peninsula, Escudero et al. (2005) reported 15% of DD intrusions (16 episodes and 54 days per year) in the period 1996-2002, and Pey et al. (2013a) obtained 17-18% between 2001 and 2011. In the comparison with our results, we must note that not all DD episodes have a net impact on PMx levels because of the mentioned characteristics of transport and deposition of DD intrusions. The contribution of these episodes to PMx and AOD will be analysed in the next subsection.

4.1.2 Annual cycle of the number of episodes and days

The annual cycles of the number of episodes and number of days with DD conditions are presented in Figure 2. In general, the seasonal pattern along the year followed by the number of episodes (Figure 2a) and dusty days (Figure 2b) is similar with a significant increase of the DD occurrence in March (13 events and 34 dusty days), a weak fall of DD event days in April (14 events and 25 dusty days), a notable increment between May and September (around 17 events and 53 dusty days per month), and a progressive decline to the minima in November and December. The number of episodes and event days peaks in June (20 events and 65 dusty days). A noteworthy feature of this figure is the unexpected local minimum in the number of DD episodes in July (15 episodes in 2003-2014) which is shifted to August in the number of DD event days (46 dusty days in 2003-2014). Figure 2b is similar in shape to that reported by Salvador et al., (2013) with the exception of September and also similar to that of Escudero et al. (2005) with the exception of October. Concerning the two types of DD conditions distinguished in our inventory, D type controls the annual cycle in the March maximum and April minimum, while MD controls the evolution between August and October.

Some features mentioned above regarding the seasonal behaviour of DD events for the North-central Spanish region are also observed for other areas of the IP. For instance, the March maximum and April minimum are common features in South-western (Toledano et al., 2007b; Obregón et al., 2012), North-eastern (Escudero et al., 2005; Papayannis et al., 2008), and Central-central (Salvador et al., 2013) Spain. In spite of the different measuring instruments (sun-photometer, lidar, or particulate mass concentration), time periods, and methodologies employed in the DD event identification, its impact over (almost) all the Iberian Peninsula seems to follow the same pattern with the two maxima, one at late-winter/early-spring (March) and the other in summer, and the accentuated minimum of winter. Several minor discrepancies are found for the rest of the year; for example, the maximum number of DD event days during summer months with a local minimum (in July) between them seems to be a characteristic of the annual cycles of the South-western (Toledano et al., 2007b) and North-central areas. Conversely, the eastern region does not show this behaviour and presents a local maximum in October (Escudero et al., 2005; Pey et al., 2013a) which is not seen in our study region. These results confirm that different areas have different aerosol properties in the IP (Mateos et al., 2015). The larger occurrence of dusty days in summer and in certain spring and fall months is also observed in other Mediterranean areas using lidar networks to identify the occurrence of DD outbreaks (e.g., Mona et al., 2006; Papayannis et al., 2008).
4.1.3 Inter-annual variability and trends of the number of episodes and days with DD

The year-to-year variability for both number of episodes and days is reported in Table 2 and illustrated in Figure 3. A large inter-annual variability is observed, more accentuated for the evaluation of the number of DD event days (Figure 3b) with an apparent decreasing trend during the analysed period. A large number of dusty days is reported during the first five years (2003-2007), the maximum being in 2006 (68 days). Beyond 2007 there is a decline of DD event days up to 19 days in 2010, and a small upturn is observed in 2011, 2012, and 2014 with a sharp reduction in 2013. The lowest occurrence of DD events occurred in 2013 with only 7 episodes and lasting 15 dusty days that year (4.11%). By contrast, the largest number of events took place in 2016 with 17 episodes composed of 68 days of dust intrusion (18.64%). Nevertheless, the number of episodes and the number of days are not directly linked. For instance, 12 episodes are observed in both 2007 and 2010 but the former 10 registered 44 days of intrusion whereas the latter just 19 days. Furthermore, even though 2006 is the year with the highest occurrence of dusty days, it is not linked with the most intense events. Both AOD and PM means (for DD days) are lower in 2006 than for previous years, in which a smaller occurrence of DD conditions is observed. The minimum load during DD days is registered in 2013 for the AOD (0.18) and in 2009 for the PM (16 µg m\(^{-3}\)), while the maximum occurred in 2004 (0.33 for AOD and 30 µg m\(^{-3}\) for PM). Concerning the two classifications of DD event days (D and MD types), years 2003, 2006, 2009, 2012, and 2014 are governed by the purer D type intrusions.

To quantify the decreasing trends in the number of episodes and associated days, the Theil-Sen estimator and Mann-Kendall test for significance have been used. The trends for the number of DD episodes and the number of days are reported in Table 3 for the yearly values. A statistically significant trend at the 95% significance level presents a p-value below 0.05 (e.g., Sanchez-Lorenzo et al., 2013). The total number of dusty days has decreased by 2.7 days per year (p-value of 0.02) between 2003 and 2014. This strongchange, however, does not cause a significant trend in the number of episodes which presents a rate decrease of 0.67 episodes per year with a p-value around 0.03 (~97% of significance level). These figures corroborate a notable significant decrease in the DD events seen in the North-central area of the Iberian Peninsula over the past decade. This result is in line with the findings obtained by Gkikas et al. (2013) for the whole Mediterranean Basin using MODIS data between 2000 and 2007 and considering only very intense DD events selected by a high AOD threshold.

4.2 Desert dust contribution to total AOD: seasonal cycle, inter-annual variability and trends

4.2.1 Annual seasonal cycle

Figure 4 and Table 4 show the annual cycle of the DD contribution (small red bars in the figure) together with the multi-annual monthly means considering all days and only days without DD aerosols (the difference between these two values gives the DD contribution). Overall, the mean DD contribution to AOD is 0.015 or 11.5% in 2003-2014.
The total AOD annual cycle representing the climatology follows the well-known pattern previously reported and explained for the Palencia site (see, e.g., Bennouna et al., 2013; Mateos et al., 2014a). To summarize: the increasing values from January to June (just where when the maximum is found), with a slight reduction in May and a decreasing trend to the end of the year, provide almost a well-defined bell shape. Concerning the climatology with the DD episodes excluded, it preserves the bell pattern found before for the general case, except for some minor discrepancies. For instance, the change between May and June is not noticeable for the curve with the DD excluded, in contrast with the larger increment observed for the general case.

However, the seasonal pattern followed by the DD contribution to AOD is considerably different to these two latter curves. Two maxima are observed during the annual cycle: the first one in March (late winter/early spring) with 0.018 or 13.4% and the strongest one occurring in summer period (June and August), ~0.027 or ~17%. Together with these maxima, there are two local minima: in April-May (around 0.014 or 9.5%) and in July (0.018 or 12%). After August, a progressive decline of the DD contribution is observed with the minimum in winter (December and January show similar values about 0.004 or 5.4%). It is worth mentioning here the different characters of the two local minima occurring in April-May and July, the former is more general of the IP (linked to the precipitation cycle) while the latter is more typical of the Central-central and South-western areas of Spain. For instance, the July minimum seems to be related with the arrival of drier air masses in the low troposphere as it is observed in the precipitable water vapour cycle (Ortiz de Galisteo et al., 2013).

The annual cycle of the DD contribution for the Palencia site (representing North-central Spain) presents a similar shape to that obtained in the “El Arenosillo” site (South-western area) by Toledano et al., (2007b) for an inventory of 6 years, from 2000 to 2005. This is an important result in two aspects, one related with the shape of the annual cycle or seasonal behaviour and the other one related with the different contribution of North-north and South-south areas of the IP. In relation to the geographical gradient of African dust contribution, a quantitative difference is observed between these two areas. The total AOD signal is clearly impacted by DD events in the Southern-southern Iberian coast (relative contributions being over the 30%), while in the North-north-central region the DD influence is weaker, thus a South-south-North-north decreasing gradient over the IP is observed regarding the DD contribution to AOD values. This behaviour is well known in the IP by earlier aerosol studies based on PMx data (Querol et al., 2009; Pey et al., 2013a; Salvador et al., 2013, 2014) but this is the first time this is confirmed by an inventory of AOD data.

4.2.2 Inter-annual variability and trends

With respect to the inter-annual change of the DD contribution to AOD, Figure 5 and Table 5 show its annual values between 2003 and 2014 (using the methodology explained in Section 3.2). In a quick-look analysis, both total AOD and DD
contributions have a significant year-to-year variability with a decreasing trend during the period studied, but with different patterns (also observed in the relative DD contribution to AOD). The maximum DD contribution with a value of 0.033 or 21.2% took place in 2004, with also a maximum in the total AOD around 0.15 (the mean value of 2003 is clearly impacted by the low sampling: 42.7%, compared to the 72.4% in 2004). The year 2013 presents the absolute minimum of the DD contribution to AOD with 0.004 or ~4%, with a low contribution in 2009 too (0.006 or ~5%). There is a weak evolution of DD contribution until 2008, although 2005 presents a marked local minimum (DD contribution to AOD around 0.016 or 11%). There are years with simultaneous decreases (2008, 2009, 2013) or increases (2011, 2014) of both total AOD and its DD contribution, but in other years they present the opposite behaviours (2005 and 2006). The solid line in Figure 5 illustrating the evolution of the relative DD contribution to AOD highlights the minima of 2013, 2009, and 2005 and the maxima of 2004 and 2012. The high inter-annual variability can be explained by the typical variability of the different African source-areas and associated emission processes together with the atmospheric conditions and transport patterns of DD aerosols that can reach the Iberian Peninsula (Prospero et al. 2002; Kaufman et al., 2005; Escudero et al., 2006; Knippertz and Todd, 2012; Salvador et al., 2014).

The temporal trends in total AOD and in the DD contribution to AOD are also evaluated and shown in Table 3. The decrease of the total AOD in the Palencia site in 2003-2014 is ~0.006 AOD-units per year (with a p-value <0.01) or ~4.6% per year, which is in line with previous findings for the same site by Bennouna et al., (2014) and Mateos et al., (2014b) for shorter (4 and 3 years, respectively) periods. With respect to the DD contribution to AOD, a rate decrease of ~0.0019 AOD-unit per year (p-value = 0.02) or ~11.2% per year is calculated. Therefore, this rate represents the 30% of the total AOD decreasing trend. Hence, the natural decrease of DD aerosols has notably affected AOD levels over North-central Iberian Peninsula during the study period.

4.3 Desert dust contribution to PMx levels: annual cycle, inter-annual variability and trends

4.3.1 Annual seasonal cycle

In the same way as for AOD, the contribution of desert dust events to mean values of PM\textsubscript{10}, PM\textsubscript{2.5}, and PM\textsubscript{2.5-10} have also been calculated. The annual cycle and the inter-annual evolution of these three quantities and the corresponding DD contributions are reported in Tables 4 and 5 and also illustrated by Figures 6 and 7, respectively.

The DD contribution to the total PM\textsubscript{10}, PM\textsubscript{2.5}, and PM\textsubscript{2.5-10} is not usually evaluated at the same time. To our knowledge, this is the first time that fine and coarse mode contributions are evaluated in a long-term desert dust inventory of this type. Furthermore, the temporal trends for the inter-annual DD contributions are also discussed. It is worth mentioning here that as PM\textsubscript{10} and PM\textsubscript{2.5} are obtained from different independent filters (see Section 2.2) while PM\textsubscript{2.5-10} is only available with
simultaneous PMx data, the \textit{data amount of datanumber} used in the evaluation of DD contribution for each quantity slightly differs.

According to Table 4, the mean DD contributions to PMx during the study period are 1.3 \(\mu g\ m^{-3}\) (12\%) for PM\(_{10}\), 0.6 \(\mu g\ m^{-3}\) (9\%) for PM\(_{2.5}\), and 0.8 \(\mu g\ m^{-3}\) (16\%) for PM\(_{2.5-10}\), respectively. Our findings during 2003-2014 are in line with those given by Querol et al., (2009): 2 \(\mu g\ m^{-3}\) for a 3-year period (2004-2006) of PM\(_{10}\) data at the Peñausende site. A decreasing south to north gradient of African dust contribution to PM\(_{10}\) (e.g., Querol et al., 2009; Pey et al., 2013a) is found for the Northnorth-central area of the IP. In particular, PM\(_{10}\) is similar to the averages in the Northnorth-eastern area (< 2 \(\mu g\ m^{-3}\)) and smaller than the values obtained in southern sites (up to 5-6 \(\mu g\ m^{-3}\)). Our relative contribution is in line with the lowest values of the ranges reported by Salvador et al. (2013) using a chemical speciation analysis in three different sites in sites near Madrid.

The total PM\(_{10}\) annual cycle (see Figure 6) is well known in the Northnorth-central area of the Iberian Peninsula (see, e.g., Bennouna et al., 2014, Mateos et al., 2015): there are two maxima, a major one in summer and a secondary one in early spring (considering our seasonal classification with March as part of the spring), a winter minimum and another minimum in April. This general behaviour for the entire dataset is also followed if the DD events are excluded. The evolution of these two latter curves is also followed by Furthermore, the evolution of the DD contribution to PM\(_{10}\) is very similar to these two latter curves. The largest DD contribution is observed in March (2.2 \(\mu g\ m^{-3}\) or 20\%) and summer months, June to August (~2.3 \(\mu g\ m^{-3}\) or ~17\%). The months of April and May (~0.9 \(\mu g\ m^{-3}\) or ~9\%) display a notable decrease with respect to March. After summer, there is a sharp fall in September (1.2 \(\mu g\ m^{-3}\) or 10\%) producing a local minimum, and beyond October a progressive decline leading to the weakest effect (<8\%) of the African intrusions during winter months (DJF). The maximum relative DD contribution to PM\(_{10}\) can reach 20\%, which is within the range (10\%-50\%) observed by Pey et al. (2013a) for the eastern Spanish coast. Comparing the seasonal cycles of DD contribution to PM\(_{10}\) in the latter area with respect to Northnorth-central Iberian Peninsula, some common features appear (March maximum, April/May decrease, summer increase, and September drop) but the maximum in October seen in the Mediterranean coast does not happen in the north-central area. A particular difference occurs in October since in the Mediterranean coast there is a notable rise of DD contribution at the surface. Even though both AOD and PM\(_{10}\) express the aerosol load, these quantities present noticeable differences. To facilitate the comparison of the results shown above, Figure S1 (supplementary information) shows together the annual cycles of AOD and PM\(_{10}\) total means and their DD contributions. The annual cycle of the two quantities, total AOD and PM\(_{10}\), for the complete dataset follows a similar behaviour between August and March, with the differences in April (local PM-minimum) and May (local AOD-minimum) being remarkable, and a slight different evolution in June-July. These discrepancies between these quantities lead to a moderate-high correlation coefficient of 0.82 between AOD and PM\(_{10}\), but their physical meaning is uncertain taking into account the mentioned discrepancies in the two annual cycles. With respect to the correlation between the seasonal cycles of DD contributions (both absolute and relative), the absolute and relative ones for AOD and PM\(_{10}\) differ show the most significant discrepancies in July (with a local minimum of AOD) and September (sharp fall of PM\(_{10}\)). Furthermore, it is worth mentioning to note that the maximum of March is more intense for the DD
contribution to PM$_{10}$ than to AOD. Hence, the correlation factors between these quantities are moderate-high: 0.84 and 0.74 for the absolute and relative curves, respectively.

The fine mode, represented by the PM$_{2.5}$ data, follows the same pattern as PM$_{10}$ in the total and DD contribution curves (Table 4 and Figure 6b). The DD contribution to PM$_{2.5}$ is below 10% for most of the year, with a mean value of ~9%.

The total coarse mode (PM$_{2.5-10}$) curve is also similar to that obtained for the total PM$_{10}$, although the mean contribution of the DD events is 16% of the total PM$_{2.5-10}$, faced to the 12% of the PM$_{10}$ (see Table 4). The DD contribution to PM$_{2.5-10}$ (Table 4 and Figure 6c) exhibits a strong maximum in March (1.7 µg m$^{-3}$ or 33%), a reduction in April and May (around 14%), large values in June (1.4 µg m$^{-3}$ or 25%) followed by a weak decrease in July and August (1.3 µg m$^{-3}$ or 21%), and low values in autumn and winter.

4.3.2 Inter-annual variability and trends

The inter-annual variations of total PM$_{10}$, PM$_{2.5}$ and PM$_{2.5-10}$ and the corresponding DD contributions to these PMx concentrations are plotted in Figure 7 and reported in Table 5. In the shape of the DD contribution we can distinguish two periods associated with the strong minimum of 2009. The first period has a decreasing trend from 2003-2009 where the first four years have similar DD contributions among them. The second period starts with a strong ascent of DD contribution from 2009 to 2012, followed by a significant fall in 2013 and a final rise in 2014. The absolute maximum DD contribution occurs in 2006 (2.4 µg m$^{-3}$ or 21%) and the absolute minimum is observed in 2013 with 0.4 µg m$^{-3}$ or 5%, although very similar to the value in 2009. The solid line in Figure 7 illustrating the evolution of the relative contribution highlights the minima of 2005, 2009 and 2013 and the maxima of 2004, 2006, 2012, and 2014.

The inter-annual evolutions of the total PM$_{10}$ and AOD are very similar (see Figure S2, supplementary information) with an excellent agreement between them represented by a correlation coefficient around 0.9 in 2003-2014. With respect to the yearly values of DD contributions to AOD and PM$_{10}$, they show a correlation coefficient of 0.81. The agreement is also quite good for the relative DD contributions to AOD and PM$_{10}$ (correlation coefficient around 0.7). This high agreement, extremely good during 2009-2013, is not seen for some years. For instance, the reason behind the low DD contribution to AOD in 2006 can be explained by the poor sampling during that year (see Table 1). So far, no reasonable explanation has been found for the strong fall between 2004 and 2005 in the DD contribution to AOD despite the fact that total AOD and PM$_{10}$ display the same behaviour. The DD contribution to PM$_{10}$ is notably larger than that obtained for AOD in 2014. The high inter-annual variability of these quantities highlights the necessity of longer time periods to assess this kind of relationships, but bearing in mind that the net contribution of DD aerosols is represented by very low values with a high uncertainty, hence this variability is within the expected range of change. These results are of extraordinary interest for long-term studies of columnar and surface aerosol loads in relation to their evolution and trends for climate studies because tropospheric aerosols have a strong regional signature and the area studied presents exceptional background conditions representative of the Western Mediterranean Basin.
The weak impact of the DD events on the PM\textsubscript{2.5} levels (fine mode, see Figure 7b and Table 5) is reflected in the low relative contribution with only three years (2003, 2005, and 2006) presenting values higher than 12%. The last years of the period analysed (2009-2014) present a notable low DD contribution to PM\textsubscript{2.5} below 7%. On the contrary, PM\textsubscript{2.5-10} (Figure 7c and Table 5) still follows presents a sharper behaviour than previous PM\textsubscript{x} results although still following the PM\textsubscript{10} pattern. The starting initial years are the ones with the largest contributions (around 27% until 2006) while 2013 shows the minimum values (around 5%) together with 2009 (~7%).

There is a decreasing trend of all the quantities shown in Figure 7. The general decrease of PM\textsubscript{x} levels has been previously reported for the Peñausende site and for shorter periods (e.g., Barmpadimos et al., 2012; Bennouna et al., 2014; Mateos et al., 2014b; Querol et al., 2014) and it has been corroborated with the temporal trends obtained in this study (see Table 3). Cusack et al., (2012) pointed out a percentage reduction ranging between 7% to 41% in the yearly PM\textsubscript{2.5} from 2002 and 2010 in 11 Spanish sites. In order to quantify the observed decrease in the DD impact, Table 3 also presents the temporal trends of the DD contribution of PM\textsubscript{10}, PM\textsubscript{2.5}, and PM\textsubscript{2.5-10}. The general decrease of PM\textsubscript{10} (-0.46 µg m\textsuperscript{-3} per year, with a p-value <0.01) in Peñausende site for the period 2003-2014 is in line with previous studies (e.g., Querol et al., 2014; Mateos et al., 2015). Regarding the DD contribution, the fall in the three quantities is quantified as around -10% per year. In particular, the DD contribution to PM\textsubscript{10} has decreased by an absolute amount of 0.14 µg m\textsuperscript{-3} per year (p-value of 0.06) and 0.08 µg m\textsuperscript{-3} per year (p-value < 0.01) for PM\textsubscript{2.5}. The reduction observed in the DD event days (see subsection 4.1.3) has also led to a significant fall of the total particulate matter. Comparing the temporal trends of the DD contribution and the rate for the total quantity, the DD impact has caused 30% of the total PM\textsubscript{10} decrease in North north-central Spain. As expected, this percentage is smaller (about 21%) for the PM\textsubscript{2.5} case. In the North north-eastern region, Querol et al., (2014) showed that crustal matter accounted for 14% of the total PM\textsubscript{2.5} decrease between 2001 and 2012.

### 4.4 Estimation of associated uncertainty of the methodology

No quantification has been done about the associated uncertainties in the number of events and associated days in most of the reported bibliography. The same happens for the uncertainty linked to the DD contribution, which can be evaluated as a consequence of the earlier error of DD detection, but also can be evaluated based on other assumptions. One way to estimate a possible range of the real uncertainty is the comparison of results obtained from different methodologies. This task was addressed by the above mentioned study of Viana et al. (2010) showing relative differences in the number of dusty days about 12% and 28-50% for the DD absolute contribution. A big step took place when the proposed methodology by Escudero et al., (2007) was taken as the official standard method. However, the 30 days moving percentile used to establish the regional background has been changed from 30% (reported by Escudero et al., 2007) to 40% (Pey et al., 2013a; Salvador et al., 2013, 2014). It seems apparent that this percentile may be site dependent thus demonstrating the difficulty of this evaluation. Otherwise, it must be borne in mind that a big difference exists between the Escudero et al., (2007) methodology and that applied by us. This subsection describes a first attempt to estimate the uncertainty associated with the method used in our study.
Fingerprints of each DD event day are visible on at least one of the quantities related to aerosol load (columnar or surface) analysed in the inventory evaluation (see Section 3), plus the additional informational of air mass back-trajectories, satellite images, and synoptic scenarios. Usually, several of these variables simultaneously corroborate the DD presence, especially due to the low background values that characterize our the north-central Spanish region. Therefore, the thorough inspection of all the information provided by different sources at the same time causes the error in the DD identification to be minimal.

From our experience during these 12 years of data, we consider that possible error sources can be, mainly, the following: gaps in the data series, classification or not of a day when the aerosol load is close to the threshold values, and uncertainty of the instrumental techniques and the ancillary tools. Therefore, we can estimate that about 3-5 days per year could be missed in the annual sum of dusty days. This assumption is based on our long-term expertise in this evaluation when the DD inventory has been re-evaluated to ensure its accuracy. So the associated relative uncertainty, considering the average of 35 DD event days per year, is ~9-15%, which is in line with the results reported by Viana et al. (2010) as mentioned above. This estimation gives a realistic range for the error associated with this methodology of visual inspection. The 5 days per year uncertainty (or 15%) can overestimate the real error, but even this percentage can be considered as acceptable as the maximum average error. Regarding the sum of dusty days in the seasonal cycle, the same range of error can be assumed in every monthly inter-annual value.

The possible possibility of missing these few days with DD fingerprints (~3-5 per year and per inter-annual month) leads to an uncertainty in the evaluation of the DD contribution to AOD values. Hence, to quantify the uncertainty in the seasonal cycle of the DD contribution to AOD each inter-annual monthly database is extended adding 9% of DD event days (considering that 3 days in are missed). For these “extra” days the AOD is assumed as the mean value during the DD events in that month. For instance, four days are added in June with a mean AOD of 0.27 and one day is added in January with AOD$_{440nm}$ = 0.18. The DD contribution is calculated for the new data series, evaluating the differences with those values shown in Section 4.2 (from the original database). The results show a small change in the DD contribution to AOD, always below 0.002. For instance, for June the relative uncertainty caused by the added days is 6.7% (the absolute DD contribution for the original evaluation is 0.027). However, those months with less absolute DD contribution to AOD cause a relative difference between 15% and 20% (such as January and December). Overall, the mean uncertainty is 0.0013 or 9.7% when the uncertainties of the twelve multi-annual monthly values are averaged. This relative uncertainty is in line with the 10% calculated by receptor modelling studies (Viana et al., 2010). The same procedure is applied for the inter-annual DD contribution to AOD. On average, the inclusion of 9% DD extra days causes an uncertainty of 0.0014 or 8.3%. If the assumption of missing 3 days per year is even enlarged increased to 5 days per year, the uncertainties caused on the DD contribution to AOD values only increase up to 14%. Hence, the reliability of the method followed here is demonstrated because of the low changes of the results when the DD inventory is added with possible dusty days missed.

In the same way, the study of the uncertainties of the DD contribution to PM$_{10}$ is also addressed with the same method (adding 9-15% extra DD event days). The results for PM$_{10}$ indicate a mean uncertainty of 0.1-0.13 µg m$^{-3}$ or 8-14% in the
evaluation of both annual cycle and inter-annual evolution. Hence, this relative uncertainty can be also extrapolated to the PM$_{2.5}$ and PM$_{2.5-10}$ DD contributions proving the feasibility of this method.

4.5 Analysis of the synoptic scenarios during desert dust episodes

Using the ancillary information used in the final choice of the DD identification, the synoptic scenarios that favour the arrival of air masses originating in the north of Africa are also studied. These scenarios are those defined and described by Escudero et al. (2005): via the Atlantic arch (North Africa High Located at Surface Level, NAH-S), directly from North-Africa by a deep low pressure (Atlantic Depression, AD) or by a convective system (North African High located at upper levels, NAH-A), and from the Mediterranean area (North African Depression, NAD). Overall, the geographical positions and heights of the high and low pressure systems produce the mineral aerosols to reach the IP. Figure 8 presents the annual cycle and inter-annual variability of the number of episodes associated with each synoptic scenario. The synoptic scenario of each episode has been established considering all the daily meteorological maps during the episode. The synoptic scenario analysis of the DD events (see Figure 8a) has shown a predominance of the NAH-A (81 out of 152 episodes), in particular, during the warm season (from May to October). This scenario corresponds to a North African High Located at Upper Levels, produced by an intense solar heating of the Saharan desert. These air masses present large DD loads which can arrive at high altitudes (up to 5 km a.s.l.). In our study region, the NAH-S (North Africa High Located at Surface Level) scenario governs (38 out of 152 episodes) the DD intrusions between December and April (being also significant in October) and produces transport in the lower atmospheric levels (generally below 1 km a.s.l.). The AD scenario (Atlantic Depression) plays a minor role (24 out of 152 episodes) but with an influence confined between February and May, September, and November. The NAD (North African Depression) scenario only presents an important contribution in March and December (9 out of 152 episodes).

The fingerprints of the evolution of these synoptic scenarios are reflected in the climatology of the DD episodes shown in Figure 2. The rapid increase in DD events in March (see Figure 2) is caused by a larger influence of NAH-S (3 to 5 DD events with respect to February), the marked appearance of NAD (3 events), and a slight increase of AD (2 to 3 DD events with respect to February). The synoptic situation in April changes and the NAD scenario almost disappears while NAH-S and AD increase their influence. The local summer minimum in July is caused by the lower occurrence of the NAH-A conditions. Previous studies have found this minimum for other columnar quantities, such as the vertical precipitable water vapour (Ortiz de Galisteo et al., 2013). The absolute DD event minimum of November is caused by the total disappearance of the NAH-A scenario.

Comparing these results with previous inventories performed in other geographical areas of the IP, the synoptic scenario climatology presents some discrepancies. Toledano et al. (2007b) have also found for the “El Arenosillo” site (Southwestern IP) in the period 2000-2005 a predominance of the NAH-A conditions during summer. However, the role played by the NAH-S seems to be minor during winter compared to the North-central area. The DD inventory in the North-Mediterranean Spanish coast has been analysed by Escudero et al. (2005) between 1996 and 2002. They also obtained the
major predominance of the NAH-A during summer, although the NAD scenario shows a notable impact on the DD events in May and November. These outbreaks arriving from the Mediterranean area are also reported in the months of February, March, and November in the “El Arenosillo” inventory.

Inter-annual distribution of DD events and the four synoptic scenarios (see Figure 8b) corroborates the predominance of the synoptic scenario NAH-A every year. Overall, there is a mean of 7 episodes per year due to this scenario in the North-central area of the IP, being the maximum influence in 2012 where 9 out 12 events occurred under this situation. A special feature is the simultaneous appearance of the four scenarios only in years 2004, 2006, and 2014. The last two years of the analysed period (2013-2014) have shown a decrease of the number of episodes that can be attributed to the absence of synoptic conditions favouring mineral dust transport during summer (NAH-A scenario). The occurrence of the NAH-S and AD scenarios presents high inter-annual variability but the number of DD episodes they caused is always smaller than those caused by NAH-A. Finally, NAD conditions in the north-central IP region are only relevant in 2004, 2011, and 2014 with 2, 3, and 2 events, respectively. However, this scenario plays a key role in the north-eastern area of the IP (e.g., Escudero et al., 2005), which shows that DD intrusions arriving through the Mediterranean area rarely reach the north-central region of Spain.

5 Conclusions

In this study, a methodology to obtain a reliable identification of DD intrusions is proposed and applied to the North-central area of the Iberian Peninsula. Long-term datasets of AOD and PMx for the background sites of Palencia and Peñausende (representative of the study area) have been used as core information for the detection of desert dust intrusions in this area during an 12-year period (from January 2003 to December 2014). The analysis of ancillary information, such as air mass back-trajectories at three altitude levels (500, 1500 and 3000 m a.s.l.), MODIS-AOD and true colour images, and meteorological maps, has been used to precisely establish the duration of each desert dust episode, creating a reliable inventory of desert dust episodes. The main conclusions can be summarized as follows:

1. The simultaneous consideration of surface and columnar aerosols has been shown to be a reliable tool in the DD identification. More than a half of the inventory has been detected by AOD and PM10 data at the same time. However, each quantity can provide the DD detection by itself in a large number of cases (114 and 80 out of 419 days detected by only PM10 and AOD data, respectively). The smaller coverage of AOD sampling is not a major handicap in this process.

2. A total of 152 episodes composed of 419 days presented desert dust aerosols during the entire period. The annual cycles of the number of DD episodes and days follow a similar pattern: an increase in March, a weak fall of event days in April, a notable increment between May and September and a progressive decline to the absolute minimum in winter, with the absolute maximum in June and local minimum in July/August. Inter-annual
variability of the number of DD episodes and dusty days is high, ranging between 7 episodes (15 dusty days) in 2013 and 17 episodes (68 dusty days) in 2006. A temporal trend of -2.7 dusty days per year (95% significance level) points out its decrease between 2003 and 2014 is obtained. Therefore, a reduction of the DD outbreaks in the North-central area of the Iberian Peninsula is found during the period studied.

3. Overall, the mean DD contribution to AOD$_{440nm}$ is 0.015 or 11.5%, while for the surface concentration PM$_{10}$, PM$_{2.5}$ and PM$_{2.5-10}$ this is 1.3 µg m$^{-3}$ (11.8%), 0.55 µg m$^{-3}$ (8.5) and 0.79 µg m$^{-3}$ (16.1%), respectively.

4. The annual cycle of the DD contribution to aerosol load peaks in March, decreases in April-May, notably increases during summer months (the AOD curve has a local minimum in July), and experiences a progressive decline after summer (with a significant fall in September for the PM$_{10}$ curve) towards minimum values in winter. The maximum DD contribution to AOD occurs in June and August close to 0.03, while the PM$_{10}$ maximum DD contribution reaches ~2.4 µg m$^{-3}$ in August.

5. The inter-annual variability of the DD contribution to aerosol load is maximum in 2004 for AOD with 0.03 and 2006 for PM$_{10}$ with 2.4 µg m$^{-3}$, and minimum in 2013 (0.004 for AOD$_{440nm}$ and 0.4 µg m$^{-3}$ for PM$_{10}$). The correlation coefficient between the DD contribution to AOD$_{440nm}$ and PM$_{10}$ yearly means is 0.81.

6. The temporal trends of the DD contribution to AOD, PM$_{10}$, and PM$_{2.5}$ have values of -0.0019 (p-value of 0.02), -0.14 µg m$^{-3}$ (p-value of 0.06) per year, and -0.08 µg m$^{-3}$ (p-value < 0.01) per year in the analysed period, respectively. All these negative rates indicate a This decrease of the levels of natural mineral dust aerosols, which represents around the 30% of the total aerosol load decrease shown by AOD (columnar) and PM$_{10}$ (surface) in 2003-2014. This decrease is around 20% for the PM$_{2.5}$ case.

7. DD outbreaks have mainly reached the North-central Iberian Peninsula directly from North-Africa by a convective system (NAH-A synoptic scenario), with clear predominance in the summer months. The NAH-S (via the Atlantic arch) and AD (directly from North-Africa by a deep low pressure) scenarios present a variable influence thorough the year, while the NAD (from the Mediterranean area) conditions are only important in March and December.

The proposed inventory is the first one based on long-term AOD-PM data series. The use of worldwide networks (EMEP and AERONET) ensures that this method can be implemented in other regions with background aerosol observations, as long as nearby PMx and AOD measurement sites in clear remote (background) locations are analysed.

With careful inspection of all the information, the inventory can be a useful tool to develop and validate automated methodologies which use other instruments such as Raman lidars and ceilometers or use model forecasts. The comparison between different methodologies will allows a more reliable estimation of uncertainties in DD detection and its contribution to total aerosol load. Future studies based on this inventory will be focused on a global characterization of microphysical and radiative properties of desert dust including the evaluation of its radiative forcing over the study region. Therefore, these results are useful for assessing regional climate change studies linked to atmospheric aerosols because of the excellent clean
background conditions of the area, which may be considered as one of the few sites/areas in Southwestern Europe with these conditions.

Acknowledgements

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References


**Tables**

Table 1: Yearly Annual sum of days with aerosol data sampling of AERONET and EMEP databases (all days: All Days) used in this study. Yearly number of dusty days in the DD inventory (desert Desert dDays) identified by criteria of AOD, PM<sub>10</sub>, AOD and PM<sub>10</sub> and other ancillary information. The relative coverage (percentage) is also given in parenthesis. See Section 3 for further details about the criteria used.

<table>
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<tr>
<th>Year</th>
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<th>2004</th>
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<th>2006</th>
<th>2007</th>
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<td>295 (80.8)</td>
<td>190 (52.1)</td>
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<td>280 (76.5)</td>
<td>256 (70.1)</td>
<td>244 (66.8)</td>
<td>269 (73.7)</td>
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<td>249 (68.2)</td>
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<td>279 (76.4)</td>
<td>183 (50.1)</td>
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<td>219 (59.8)</td>
<td>200 (57.5)</td>
<td>234 (64.1)</td>
<td>2771 (63.2)</td>
</tr>
</tbody>
</table>

| Desert Dust Event Days |      |      |      |      |      |      |      |      |      |      |      |      |       |
| Number of dusty days | 44 | 44 | 41 | 6867 | 44 | 31 | 24 | 19 | 32 | 29 | 15 | 28 | 419418 |
| Only AOD Criterion (%) | 5 (11.4) | 8 (18.2) | 9 (22.0) | 6 (18.9) | 14 (31.8) | 4 (12.9) | 4 (16.7) | 5 (26.3) | 9 (28.1) | 12 (41.4) | 2 (13.3) | 2 (7.1) | 80 (19.1) |
| Only PM<sub>10</sub> Criterion (%) | 19 (43.2) | 3 (6.8) | 11 (26.8) | 37 (84.9) | 6 (13.6) | 2 (6.5) | 7 (29.2) | 2 (10.5) | 0 (0) | (3.4) | 8 (53.3) | 18 (64.3) | 114 (27.35) |
| AOD&PM<sub>10</sub> Criteria (%) | 20 (45.5) | 33 (75.0) | 21 (51.2) | 234 (50.0) | 22 (45.8) | 25 (80.6) | 11 (47.4) | 9 (71.9) | 23 (51.7) | 15 (26.7) | 4 (28.6) | 8 (51.42) | 2145 (51.2) |
| Other Criteria (%) | 0 (0) | 0 (0) | 0 (0) | 1 (1.5) | 2 (4.5) | 0 (0) | 2 (8.3) | 3 (15.8) | 0 (0) | 1 (3.4) | 1 (6.7) | 0 (0) | 10 (2.4) |
Table 2: Main results of the DD inventory. Legend: N.E. (number of episodes), N.D. (number of days), P.D. (Percentage of days), and M.D. (mean duration). Yearly mean values of AOD, α and PMx data of desert dust events are also reported.

<table>
<thead>
<tr>
<th>Year</th>
<th>2003</th>
<th>2004</th>
<th>2005</th>
<th>2006</th>
<th>2007</th>
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<th>2009</th>
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<th>2013</th>
<th>2014</th>
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<td>44</td>
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<td>2.14</td>
<td>3.11</td>
<td>2.7</td>
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| Mean AOD<sub>440nm</sub> | 0.32±0.11 | 0.33±0.16 | 0.28±0.13 | 0.24±0.08 | 0.29±0.11 | 0.27±0.10 | 0.20±0.05 | 0.31±0.11 | 0.27±0.11 | 0.18±0.10 | 0.19±0.09 | 0.18±0.09 | 26.0 ±0.05 |
| Mean α | 0.98±0.33 | 0.92±0.40 | 0.95±0.32 | 0.88±0.40 | 1.17±0.40 | 0.92±0.44 | 0.91±0.31 | 0.90±0.36 | 0.83±0.48 | 1.22±0.29 | 0.63±0.41 | --±0.94 | 0.99±0.15 |
| Mean PM<sub>10</sub> (µg m<sup>-3</sup>) | 28.7±13.0 | 30.0±32.7 | 29.8±28.5 | 21.2±12.0 | 19.3±12.0 | 21.5±8.0 | 16.0±6.5 | 25.0±25.8 | 21.8±11.0 | 23.2±20.4 | 16.4±8.8 | 22.9±4.7 |
| Mean PM<sub>2.5</sub> (µg m<sup>-3</sup>) | 14.9±6.3 | 14.4±8.9 | 14.7±10.0 | 12.0±3.7 | 10.0±4.1 | 13.8±5.0 | 8.5±9.4 | 10.6±3.1 | 8.7±3.3 | 7.9±3.7 | 8.5±2.6 | 11.0±2.8 |
| Mean PM<sub>2.5-10</sub> (µg m<sup>-3</sup>) | 13.9±9.1 | 15.9±25.1 | 14.7±20.9 | 9.2±6.5 | 9.3±8.4 | 7.6±5.8 | 7.6±3.1 | 14.4±16.5 | 12.5±8.1 | 15.5±17.7 | 7.6±4.7 | 14.6±8.0 | 11.9±3.4 |
Table 3: Temporal trends (Theil-Sen estimator), p-value and confidence interval ([i1, i2]) given by the quantities considered for all days and for the contribution of DD. For the DD inventory the number of episodes and DD event days are also included. **Negative trend means a decrease of the quantity analysed.**

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<th>Quantity</th>
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<th>i2</th>
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<td>-0.003</td>
<td>AOD-units per year</td>
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<td>PM$_{10}$</td>
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<tr>
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<td>-0.19</td>
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<td>Number of Episodes</td>
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<tr>
<td>DD Contribution to AOD</td>
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<td>-0.16</td>
<td>0.00</td>
<td>µg m$^{-3}$ per year</td>
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Table 4. Monthly mean (and total) contribution of DD to total AOD and PMx, in absolute (\textit{abs.}, AOD-units and \(\mu g\) m\(^{-3}\), respectively) and relative (\textit{rel.}, \%) values during the 2003-2013 period.

<table>
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<th>Jan</th>
<th>Feb</th>
<th>Mar</th>
<th>Apr</th>
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<th>Jul</th>
<th>Aug</th>
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<th>Oct</th>
<th>Nov</th>
<th>Dec</th>
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<tr>
<td><strong>AOD(_{440}) nm</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
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<td></td>
<td></td>
<td></td>
<td></td>
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<tr>
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<td>0.014</td>
<td>0.014</td>
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<td>0.018</td>
<td>0.026</td>
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<td>0.014</td>
<td>0.008</td>
<td>0.004</td>
<td>0.015</td>
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<tr>
<td><strong>PM(_{10})</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
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<td></td>
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<tr>
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<td>0.58</td>
<td>2.23</td>
<td>0.81</td>
<td>0.96</td>
<td>2.28</td>
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<td>1.37</td>
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<td>6.73</td>
<td>20.13</td>
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<td>8.54</td>
<td>17.69</td>
<td>16.51</td>
<td>16.13</td>
<td>9.61</td>
<td>13.57</td>
<td>11.57</td>
<td>3.64</td>
<td>11.80</td>
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<tr>
<td><strong>PM(_{2.5})</strong></td>
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<tr>
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<td>0.66</td>
<td>0.28</td>
<td>0.33</td>
<td>0.88</td>
<td>1.12</td>
<td>1.12</td>
<td>0.58</td>
<td>0.50</td>
<td>0.28</td>
<td>0.08</td>
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<tr>
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<td>11.99</td>
<td>13.64</td>
<td>13.40</td>
<td>8.33</td>
<td>9.71</td>
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<td>1.88</td>
<td>8.53</td>
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<tr>
<td><strong>PM(_{2.5-10})</strong></td>
<td></td>
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<tr>
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<td>0.17</td>
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<td>0.47</td>
<td>0.69</td>
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<td>1.25</td>
<td>1.33</td>
<td>0.58</td>
<td>0.93</td>
<td>0.56</td>
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<td>0.79</td>
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Table 5. Mean annual contribution of DD to total AOD and PMx in absolute (‘abs.’, AOD-units and µg m\(^{-3}\), respectively) and relative (‘rel.’, %) values during the 2003-2013 period.

<table>
<thead>
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<td>21.22</td>
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<td>14.78</td>
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<td>12.03</td>
<td>14.67</td>
<td>3.76</td>
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<td>17.80</td>
<td>16.10</td>
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<td>11.04</td>
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<td>0.68</td>
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<td>4.76</td>
<td>6.92</td>
<td>5.99</td>
<td>6.88</td>
<td>3.52</td>
<td>6.55</td>
</tr>
<tr>
<td>PM(_{2.5-10})</td>
<td>1.38</td>
<td>1.61</td>
<td>1.21</td>
<td>1.21</td>
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<td>14.42</td>
<td>18.31</td>
<td>4.88</td>
<td>22.57</td>
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Figure captions

Figure 1. Location of the main sites used in this study belonging to AERONET (blue diamonds) and EMEP (red stars) networks.

Figure 2. Annual cycle of a) total number of episodes per month for total DD intrusions; b) total (blue bars) number of days per month for total DD intrusions and for desert (D, green bars) and mixed desert (MD, red bars) categories in 2003-2014. Mean values per month can be derived dividing by 12.

Figure 3. Inter-annual variability of a) total number of episodes; b) total (blue bars) number of days for the desert dust (DD) intrusions and for desert (D, green bars) and mixed desert (MD, red bars) categories.

Figure 4. Annual cycle for DD contribution to the total monthly AOD means in absolute (red bar) and relative values (black line) in 2003-2014. Blue bars (also indicated as All Data) represent the annual cycle of total AOD and green bars the corresponding values without including the days of desert dust (indicated as Non desert).

Figure 5. Inter-annual variability of DD contribution to the total yearly AOD in absolute (red bar) and relative values (black line). Blue bars (also indicated as All Data) represent the mean year AOD value and green bars the corresponding values without including the days of desert dust.

Figure 6. Annual cycle for DD contribution to the total monthly PM$_{10}$ (a), PM$_{2.5}$ (b), and PM$_{2.5-10}$ (c) means in absolute (red bar) and relative values (black line) in 2003-2014. Blue bars represent the annual cycle of total PM$_{10}$ (a), PM$_{2.5}$ (b), and PM$_{2.5-10}$ (c) and green bars the corresponding values without including the days of desert dust.

Figure 7. Inter-annual variability of DD contribution to the total yearly PM$_{10}$ (a), PM$_{2.5}$ (b), and PM$_{2.5-10}$ (c) means in absolute (red bar) and relative values (black line). Blue bars represent the mean year PM$_{10}$ (a), PM$_{2.5}$ (b), and PM$_{2.5-10}$ (c) value and green bars the corresponding values without including the days of desert dust.

Figure 8. Annual cycle (a) and inter-annual (b) variability of DD episodes classified in terms of their synoptic scenarios: NAH-S (white bars), AD (green bars), NAD (red bars), and NAH-A (blue bars). The four synoptic scenarios are described in the text (see Section 4.5).
Figures

Figure 1

Figure 1: Location of the main sites used in this study belonging to AERONET (blue diamonds) and EMEP (red stars) networks.
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Figure 6: Annual cycle for DD contribution to the total monthly $\text{PM}_{10}$ (a), $\text{PM}_{2.5}$ (b), and $\text{PM}_{2.5-10}$ (c) means in absolute (red bar) and relative values (black line) in 2003-2014. Blue bars represent the annual cycle of total $\text{PM}_{10}$ (a), $\text{PM}_{2.5}$ (b), and $\text{PM}_{2.5-10}$ (c) and green bars the corresponding values without including the days of desert dust.
Figure 7: Inter-annual variability of DD contribution to the total yearly PM$_{10}$ (a), PM$_{2.5}$ (b), and PM$_{2.5,10}$ (c) in absolute (red bar) and relative values (black line). Blue bars represent the mean year PM$_{10}$ (a), PM$_{2.5}$ (b), and PM$_{2.5,10}$ (c) value and green bars the corresponding values without including the days of desert dust.
Figure 8: Annual cycle (a) and inter-annual (b) variability of DD episodes classified in terms of their synoptic scenarios: NAH-S (white bars), AD (green bars), NAD (red bars), and NAH-A (blue bars). The four synoptic scenarios are described in the text (see Section 4.5).