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On the aerosol weekly cycle spatiotemporal variability over Europe

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In this work, we focus on the spatial and temporal variability of the aerosol weekly cycle over Europe as these were recorded from TERRA MODIS and AQUA MODIS satellite instruments. Aerosol optical properties retrieved from MODIS TERRA (February 5 2000–February 2009) and AQUA (July 2002–December 2008) were used to produce an aerosol weekly cycle index. First, the general aerosol optical depth ($AOD_{550\text{ nm}}$) weekly patterns were defined at a $1^\circ \times 1^\circ$ resolution using the satellite-based index and six regions of interest were selected. To remove episodic dust transport events, two different aerosol flags, employing fine mode ratio ($FMR_{550\text{ nm}}$) and $AOD_{550\text{ nm}}$ data, 10 were applied diagnostically. A second spatial averaging method was then used for the investigation of the weekly variability and the statistical significance of the weekly cycle over each of the previously selected regions. Three major weekly cycle plumes are observed over Europe. A strong positive (higher values during midweek) weekly cycle plume appears over Central Europe, while a strong negative (higher values during 15 weekend) weekly plume appears over the Iberian Peninsula and the North-eastern Europe. A weak but statistically significant negative plume is apparent over the Eastern Mediterranean. The temporal examination of the weekly cycles shows that in some areas there are seasonal differences in the sign of the weekly cycle. The aerosol weekly variability over different regions in Europe was examined in conjunction with 20 the dominating synoptic wind patterns from the NCEP/NCAR reanalysis, showing that the seasonal weekly cycle plumes over regions situated in the eastern Europe and the Mediterranean Sea could be partly attributed to the westerly transport of continental aerosols.

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1 Introduction

For decades, scientists investigate the impact of human activities on atmospheric composition, meteorological and climatological parameters. The effect of human activities on these parameters and the complex connections between them cannot always be easily detected. A potent tool for the investigation of the anthropogenic effect on regional pollution, meteorological and climate variables is the identification of weekly cycles in them. Since normally there are no weekly cycles in nature, it becomes obvious that this is a strictly anthropogenic effect caused by the working cycle.

Weekly cycles of photochemical parameters, ozone, primary pollutants such as NO_x , CO and aerosols were observed from mid 70s in several urban centers (e.g. New York; Washington, DC; Los Angeles) in the US (Cleveland et al., 1974; Lonneman et al., 1974; Lebron, 1975; Elkus and Wilson, 1977) with the idea of considering weekday-weekend variations dating back to the late work of Haagen-Smit and Brunelle (1958). The years that followed, many studies particularly in America and Europe (Cleveland and McRae, 1978; Karl, 1978; Bower et al., 1989; Altshuler et al., 1995; Pryor and Steyn, 1995; Brönnimann and Neu, 1997; Vukovich, 2000; Pont and Fontan, 2001; Marr and Harley, 2002; Heuss et al., 2003; Paschalidou and Kassomenos, 2004; Shutters and Balling, 2006; Murphy et al., 2007; Stephens et al., 2008; Schipa et al., 2009) have shown that in urban areas ozone concentrations generally maximize on the weekend when emissions of NO_x and volatile organic compounds (VOCs) are lowest due to decreased vehicle emissions, while in rural areas or areas with high biogenic VOC emissions ozone may minimize on the weekend. The aerosol concentration weekend effect has also been examined for sites situated in Central and North America (Stephens et al., 2008; Murphy et al., 2008), Europe (Barmet et al., 2009) and Asia (Gong et al., 2007; Choi et al., 2008; Kim et al., 2009). The majority of the studies on aerosol concentration weekly cycles have focused on urban sites suggesting lower values during weekends and higher during workdays, which is more or less expected due to decreased industrial activity and commuter traffic during weekends. However,

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a work from Murphy et al. (2008), where measurements from both urban and rural stations around the US were used, suggested that the aerosol concentration weekly cycles are unlikely to be from local sources at the sampling sites only.

Today, there is evidence that the aerosol weekly cycle is not a local issue restricted to the ground level but it can extend over larger geographical areas in the troposphere and hence be detectable not only from ground but also from satellite-based instruments. Jin et al. (2005) using summer aerosol optical depth (AOD) measurements from the Aerosol Robotic Network (AERONET) detected a weekly variability of the columnar aerosol load over New York. Bäumer et al. (2008) using AERONET AOD_{440 nm} measurements from 14 stations in Central Europe found a significant weekly cycle for 12 of them with a mean difference between minimum and maximum of ~20% for stations in Germany and the greater Paris area and ~10% for stations in northern Italy and Switzerland. Beirle et al. (2003) initiated the effort of detecting pollution weekly cycles from space using total NO₂ measurements from Global Ozone Monitoring Experiment (GOME) instrument aboard ERS-2. They found a Sunday minimum with reduced NO₂ tropospheric columns of ~25–50% for the US, Europe and Japan. No weekly cycle was observed over China while for Middle East, Friday or Saturday (for Israel), was the day of minimum, indicative of the cultural and religious differences between these regions. Xia et al. (2008) examined the weekly cycle of the AOD_{550 nm} using satellite observations from Moderate Resolution Imaging Spectroradiometer aboard EOS TERRA (MODIS TERRA) in conjunction with ground-based data from the AERONET. The 6-year averages show a statistically significant weekly cycle in different regions around the globe (Europe and the US). An inverse phase of the weekly cycle was observed for eastern China which is in contrast with ground level observations (Gong et al., 2007). In a very recent work, Quaas et al. (2009) presented spatially averaged results using satellite observations from MODIS and data from two global climate models. They found clear aerosol weekly cycles in the data above Europe while the model runs indicated that the differences in emissions between weekdays and weekends may lead to such a cycle. In parallel with ozone, aerosols and the other atmospheric pollutants,

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there are many studies investigating the anthropogenic influence on weekly cycles in several meteorological parameters. These include rainfall amount and frequency, surface temperature, diurnal temperature range, precipitation, wind speed, etc. (e.g. Gordon, 1994; Simmonds and Keay, 1997; Cerveny and Balling Jr., 1998; Forster and Solomon, 2003; Jin et al., 2005; Bäumer and Vogel, 2007; Gong et al., 2007; Bell et al., 2008; Sanchez-Lorenzo et al., 2008; Laux and Kunstmann, 2008; Kim et al., 2009, 2010).

In this work, we have focused on the spatial and temporal variability of the aerosol weekly cycle over Europe (30° N– 70° N, 15° W– 60° E) as this was recorded from MODIS TERRA and AQUA satellite instruments. Aerosol optical properties retrieved from TERRA (February 2000–February 2009) and AQUA (July 2002–December 2008) MODIS were used in order to produce an aerosol weekly cycle index. This method was initially proposed by Xia et al. (2008); however, as will be shown here, the method has limitations which should be taken into account. The general weekly patterns were defined with the use of the satellite-based index and 6 regions of interest were selected. A second method was used for the investigation of the weekly variability and the statistical significance of the weekly cycle over the selected regions. This method was originally applied by Quaas et al. (2009) on MODIS and modeled aerosol data for a broad area including Central Europe. Here, it is shown that this kind of analysis returns better results when applied for smaller regions with uniform weekly spatial patterns. The weekly cycle was also examined on a seasonal basis. The different aerosol weekly variability over different regions in Europe was examined in conjunction with the dominating synoptic wind patterns. For this, synoptic wind speed and direction at the 850 mbar pressure level from the NCEP/NCAR reanalysis dataset were used in conjunction with the MODIS datasets.

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2 Data and methods

2.1 Data

The satellite data used here are part of the level-3 MODIS TERRA (MOD08_D3) and MODIS AQUA (MYD08_D3) $1 \times 1^\circ$ daily gridded Collection 005 dataset. The data have been acquired through LAADS (Level 1 and Atmosphere Archive and Distribution System) (<http://ladsweb.nascom.nasa.gov>). MODIS TERRA was launched on board the Terra satellite on 18 December 1999, with daytime equator crossing at 10:30 LT and MODIS AQUA was launched on board the AQUA satellite on 4 May 2002, with daytime equator crossing at 13:30 LT. MODIS with a viewing swath of 2330 km provides almost daily global coverage. MODIS measures backscattered radiation at 36 spectral bands between 0.415 and 14.235 μm with a spatial resolution of 250, 500 and 1000 m. Two separate algorithms, one for land (Kaufman et al., 1997; Levy et al., 2007a, b; Remer et al., 2005) and one for ocean (Tanré et al., 1997; Levy et al., 2003; Remer et al., 2005) are used for the retrieval of aerosol optical properties. MODIS aerosol properties from different collections have been consequently validated against AERONET Sun photometer measurements (e.g. Chu et al., 2002; Remer et al., 2002, 2005; Levy et al., 2010). The pre-launch uncertainty of the MODIS aerosol optical depth (τ) is $\pm 0.05 \pm 0.2\tau$ over land (Chu et al., 2002), and $\pm 0.03 \pm 0.05\tau$ over ocean (Remer et al., 2002). The pre-launch uncertainty of the fraction of fine mode to total aerosol optical depth or else fine mode ratio is $\pm 30\%$ over oceans (Remer et al., 2005). In this work, the aerosol optical depth at the wavelength of 550 nm over land and ocean (AOD_{550}) (`Optical_Depth_Land_And_Ocean_Mean`) and the fraction of fine mode to total aerosol optical depth or else fine mode ratio (FMR_{550}) (`Optical_Depth_Ratio_Small_Land_And_Ocean_Mean`) over both land and ocean from Collection 005 were used. In general, FMR_{550} over ocean is rather trustful; however, over land it can only give an indication of the type of aerosols (dust/non-dust) being more a qualitative rather than a quantitative measure (L. Remer, personal communication, 2010). The analyzed datasets span from February 2000 to February 2009 for TERRA

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and from July 2002 to December 2008 for AQUA MODIS. The region under investigation is the greater European area (30° N– 70° N, 15° W– 60° E) which is represented in 3000 $1 \times 1^{\circ}$ geographical cells.

Daily mean, wind speed and direction at the 850 mbar pressure level for the period 5 2000–2009 were obtained from National Center for Environmental Prediction/National Center for Atmospheric Research (NCEP/NCAR) Reanalysis data. The NCEP/NCAR Reanalysis contains global meteorological conditions with a 2.5° horizontal resolution and a 17 level vertical resolution at 6 h time intervals (Kalnay et al., 1996). The re-analysis dataset captures synoptic scale dynamic features thus being suitable for the 10 investigation of large scale aerosol transport patterns.

2.2 Data analysis

A set of previously used methods was implemented in order to obtain a balance between the detection of the spatial and temporal variability of weekly cycles and the statistical significance of our results. For the detection of the weekly patterns a method 15 similar to Xia et al. (2008) was followed. Calculations were made for each grid cell separately. This method was applied on the AOD550 data and two new datasets emerged from the initial one. For each grid cell i , a daily value ($d_{ijmy,new}$) corresponding to day j (Monday–Sunday) of week m and year y is expressed as a percentage departure (%) of the initial daily values ($d_{ijmy,old}$) from the weekly average (wa_{ijmy}) using the equation 20 presented bellow:

$$d_{ijmy,new} = 100(d_{ijmy,old} - wa_{ijmy})/wa_{ijmy} \quad (1)$$

where $j = 1–7$ and $m = 1–52$ (or 53 depending on the year). This dataset was used for the calculation of the average percent departure (APD) of each day of the week (7 values for each cell i , one for each day of the week from Monday to Sunday) from the 25 weekly average for the whole period of interest. For the investigation of the general weekly patterns a second dataset was compiled. For each grid cell i there is one value w_{ijmy} for each week m of a year y corresponding to the percentage of the difference

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between the average value of the initial dataset for Wednesday-Friday and the average value for Saturday-Monday to the weekly average $wa_{i,my}$.

$$w_{i,my} = 100 \left(\sum_{j=\text{Wed-Fri}} d_{ij,my,\text{old}} - \sum_{j=\text{Sat-Mon}} d_{ij,my,\text{old}} \right) / wa_{i,my} \quad (2)$$

By averaging the $w_{i,my}$ values for the whole period of interest one obtains the mean percentage of the difference between midweek and weekend to the weekly average for each grid cell, hereafter denoted as Weekly Cycle Index (WCI). Three-day averages were chosen instead of two-day in order to obtain more robust statistics. Only weeks that have at least two days with at least one day being among Wednesday-Friday and at least one being among Saturday–Monday were used in both datasets. The aforementioned sampling methods render systematic weekly cycles more evident, similar to the diurnal analysis of Smirnov et al. (2002), since the values are normalized with the special characteristics of each week. A two tailed t-test was used in order to examine whether the APD and WCI values are statistically different from zero. Statistical significance is indicated only when it exceeds the 90% confidence level. However, the inclusion of small negative optical depth values (≥ -0.05) in Collection 005 (Remer et al., 2006; Levy et al., 2009) could lead to false results when using WCI to examine the weekly cycle patterns. For places with very low aerosol load where small negative and positive values appear in the same week, the weekly means could be zero or very close to zero while the weekday-weekend difference on the other hand could be significantly different than zero. Since the calculation of WCI includes normalization of weekday-weekend differences $w_{i,my}$ with the weekly mean $wa_{i,my}$, those zero values could lead to huge, negative or positive, individual values which in turn could give false patterns. Several tests have been applied in order to detect anomalously high or low $w_{i,my}$ values. The limit of $\pm 1000\%$ was set to $w_{i,my}$ which proved to be helpful in “clearing” the WCI patterns. Despite the fact that fine mode ratio (FMR_{550}) expresses more a dust/non-dust flag usually taking values 0 or 1 over land (Remer et al., 2005) rather than the ratio of AOD_{550} dedicated to fine mode aerosols as it does over oceans, the

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aerosol type flag originally suggested by Barnaba and Gobbi (2004) and hereafter denoted as (flag1) may be used for the identification of cases where dust is the dominating aerosol type, thus eliminating the effect of episodic aerosol transport (e.g. from Sahara) on the signal of the weekly cycle. In cases with $AOD_{550} \geq 0.3$ and $0 \leq FMR_{550} \leq 0.7$ the dominant aerosol type is dust. A quite similar but less strict flag (including mixed type aerosols) previously used by Kaskaoutis et al. (2007) was also used (flag2). Here, in cases with $AOD_{550} \geq 0.3$ and $0 \leq FMR_{550} \leq 0.6$ the dominant aerosol type is classified as dust. After the recognition of the general weekly patterns, box regions with different weekly cycle characteristics were selected in order to generalize our results.

5 Mean values of APD and WCI for each of the box regions were calculated from the corresponding grid results following the formula presented below:

10

$$\text{Mean} = \left(\sum \text{obs}_i \text{mean}_i \right) / \left(\sum \text{obs}_i \right) \quad (3)$$

where obs_i is the number of observations used for the calculation of mean_i (APD) for each grid cell i .

15 After the definition of the regions of interest we followed a method previously presented by Quaas et al. (2009) for the investigation of the statistical significance of the weekly cycles. However, we argue that this method, should be used only after regions of interest have been identified and not for very large regions, as used originally in Quaas et al. (2009). A third dataset consisting of daily area averages for each of the 6 selected regions was created. A total of 12 time series (6 for TERRA and 6 for AQUA) were analyzed separately. The APD of each day of the week from the weekly average was calculated for each of the time series using the method outlined previously (Eq. 1). Again, only weeks with at least one day being among Wednesday–Friday and at least one being among Saturday–Monday were used. The same procedure was repeated 20 for hypothetical 6-day and 8-day day weeks. The 6 and 8-day weekly cycle was compared to the 7-day cycle to examine whether the signal in the later case is stronger or not. This gives an indication of the significance of the 7-day cycle. However, as it was discussed in Choi et al. (2008), except from the fact that the time series are influenced

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by the synoptic weather noise on approximately a few days to 14 days (a period that includes the 7-day cycle), the 7-day cycle may fluctuate by 1 or 2 days due to contamination from secondary aerosol formation and/or aerosol loading. According to this, the existence of a 7-day signal much stronger than the 6 and 8-day signal is an indication of a statistically significant local weekly cycle. On the contrary, the existence of a clear 7-day signal comparable to the 6 and 8-day signal does not necessarily undermine the statistical significance of the weekly cycle but it could be an indication of a synoptically driven weekly cycle due to transport of aerosols from other regions.

In addition, the discrete Fourier transform X of the time series was calculated by using the Fast Fourier Transform (FFT) algorithm. Despite the fact that the time series emerged after spatially averaging over much larger areas than a single $1 \times 1^\circ$ grid cell there were still days with no measurements. Like in previous studies (Hies et al., 2000; Marr and Harley, 2002), before applying the FFT a logarithmic transformation for variance stabilization was applied on the original time series. The average log concentration was extracted from all values to obtain a zero mean for the series. The missing values were set to zero. 12 new time series emerged and subsequently the periodogram was calculated using the following formula:

$$\Phi(v_k) = |X(k)|^2 = \left| \frac{1}{N} \left(\sum_{t=0}^{N-1} x_t \exp\{-2\pi i v_k t\} \right) \right|^2 \quad (4)$$

where $\Phi(v_k)$ is the squared amplitude of the spectrum (i.e. the spectral density), $k = 0, 1, \dots, N - 1$ with N being the number of the observations, x_t the new time series and $v_k = k/N$. The periodogram indicates the strength of the signal as a function of frequency and its spectra over the frequency range corresponds to the variance of the time series data. The existence of a spectral peak at a frequency equal to $1/7$ days $^{-1}$ is a strong indication of a 7-day cycle. The statistical significance of a seven day period is investigated through a red noise fit using a first-order autoregressive model AR(1) (Mann and Lees, 1996; Weedon, 2005; Wilks, 2006). A red noise background spectrum was used because geophysical time series tend to have larger power at lower

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5 frequencies (Ghil et al., 2002). The red noise spectrum is calculated from the lag-one autocorrelation coefficient of the time series and the average value of the power spectrum using Eq. (4) from Mann and Lees (1996). The significance of the spectral peaks can be estimated assuming that the uncertainties of the spectral estimates follow a
5 chi-squared distribution (Priestley, 1981; Percival and Walden, 1993). The spectral confidence intervals are calculated using Eq. (8.79) from Wilks (2006) from the degrees of freedom of the estimates (here 2) divided by the χ^2 value appropriate to the confidence interval required (here 90% or 95%). When a spectral peak exceeds the corresponding interval then statistical significance is indicated.

10 The seasonal WCI patterns were also examined defining four seasons, December–January–February (DJF), March–April–May (MAM), June–July–August (JJA) and September–October–November (SON) which are typical for the mid-latitudes. The seasonal weekly variability was examined for the 6 box regions using the APD values calculated from the MODIS TERRA and AQUA time series. The seasonal variability
15 of aerosols is examined in conjunction with daily mean synoptic wind speed and direction for the 850 mbar pressure level data from the 2000–2009 NCEP/NCAR reanalysis dataset (Kalnay et al., 1996). The wind data are available on a $2.5 \times 2.5^\circ$ spatial resolution. It is proposed here that aerosol weekly cycles could be transferred away from areas with strong and dominating weekly patterns, affecting the regional weekly patterns
20 over areas with no or weak local weekly variability.

3 Results and discussion

3.1 Detection of aerosol weekly patterns

25 The calculation of WCI for MODIS TERRA and AQUA for the period February 2000–February 2009 and July 2002–December 2008, respectively, revealed quite similar spatial patterns for the weekly cycle of aerosols across Europe (see Fig. 1a and b). 6 regions with different characteristics were studied in more detail. The 6 regions are

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marked with a thick white outline in all the maps appearing in this paper. In Fig. 1a the region examined in Quaas et al. (2009) (QE) is also marked with a thick black line. It is quite obvious that this greater region encircles regions with differing weekly cycles. The grid cells with a statistically significant weekly cycle at the 90% confidence level according to the two tailed t-test appear with a thin white outline. Central Europe (CE) is characterised by a clear positive cycle with higher values during midweek than during weekend. The positive weekly plume extends from western France to Poland and from northern Italy to northern Germany. A clear negative cycle appears over the box regions surrounding South-western Europe (SWE) which covers almost the whole Iberian Peninsula and North-eastern Europe (NEE) which is situated on the northeast of CE. In all three cases, many grid cells with statistically significant weekly cycle are situated within each box region, especially in the case of MODIS TERRA. The other three regions selected are Central Mediterranean (CM), Eastern Mediterranean (EM) and Central-Eastern Europe (CEE). They all present a moderately negative weekly cycle, with few grid cells with significant weekly cycle in the case of CM and CEE. Despite the fact that regions with a strong positive or negative weekly cycle are located within QE, this region is characterized by a slightly negative WCI. This shows that QE should not be regarded as one region as far as weekly cycle is concerned. The exact geolocation of the 6 box regions examined here, including QE, along with the corresponding average WCI values, the statistical significance and the observations used for the calculations, are given in Table 1.

As it was discussed in Sect. 2.2., the WCI could lead to false results for regions with low aerosol load due to the inclusion of small negative values in MODIS Collection 005 datasets. After applying the limit of $\pm 1000\%$ to the w_{imy} values used in the calculation of WCI the patterns were significantly “cleared” as seen in Fig. 1c and d. The corrected WCI values for the 6 regions are also cited in Table 1. Only $\sim 0.03\%$ and $\sim 0.06\%$ of the values is filtered during the correction for TERRA and AQUA which gives an indication of how few the anomalously high and low values are. The $\pm 1000\%$ limit correction proves to be proper for the kind of analysis applied here since it combines

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effectiveness in clearing the patterns and preserving the statistical sample. The WCI data presented hereafter are from the new corrected dataset. Indicatively, two typical examples of false WCI values are presented in Fig. 2a and b for MODIS TERRA and AQUA, respectively. In the case of TERRA, the grid cell centered in (37.5° N, 7.5° W) 5 has two w_{imy} values (−1151% and +10 500%) that exceed the limit of $\pm 1000\%$ with all the other values ranging from −312% to +185% (Fig. 2a). This grid cell appears as an individual red pixel in the region of SWE which is generally covered by blue-green pixels (Fig. 1a). After subtracting those two anomalously high values, the WCI from highly positive (+509%) turns to negative (−5%), which is very close to the mean value 10 for the region of SWE where the grid cell is located (Fig. 1c). The same situation stands for a neighbouring grid cell centered in (38.5° N, 6.5° W) which appears as an individual orange pixel in SWE in the corresponding AQUA MODIS WCI map (Fig. 1b). After the subtraction of two positive values (+1311% and +2759%), which exceed the $\pm 1000\%$ limit (Fig. 2b), the WCI turns from +12.90% to −1.23% (Fig. 1d). It has to be mentioned 15 that the remaining w_{imy} values range from −777% to +352% (Fig. 2b).

The impact on aerosol properties of Sahara dust episodic outbursts, which affect Europe and especially the regions around the Mediterranean Sea during spring and summer, has been thoroughly examined using satellite observations (e.g. MODIS, TOMS, SeaWiFS, etc.), ground-based observations (e.g. lidars, cimel sunphotometers, radiometers, PM ground stations, etc.) and models (e.g. DREAM, HYSPLIT, etc.) 20 (Barnaba and Gobi, 2004; Papayannis et al., 2005; Tafuro et al., 2006; Kalivitis et al., 2007; Toledano et al., 2007; Meloni et al., 2007; Gkikas et al., 2009). In this work, two aerosol flags, presented in Sect. 2.2., have been applied for diagnostic purposes 25 on the already corrected data in order to eliminate the impact of episodic Sahara dust events. The corrected with flag1 and flag2 weekly patterns can be seen on Figs. 1e and 1g for TERRA MODIS and Fig. 1f and h for AQUA MODIS, respectively. The differences between flag1 and flag2 corrected patterns are very limited. As it was expected, the general patterns did not change significantly for continental regions in Europe. On the other hand, some changes appear over the Mediterranean and the surrounding

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coastal regions. For CE and NEE there is practically no change in the average WCI for both TERRA and AQUA. The average WCI remains negative but its absolute value decreases by $\sim 38\%$ (TERRA) and $\sim 66\%$ (AQUA) in the case of SWE due to an increase of WCI for the regions in southern and southern-eastern Iberian Peninsula. This is more obvious in the case of AQUA measurements. For CM the average WCI remains close to zero, although it turns from slightly negative to slightly positive. The situation is quite similar for EM and CEE.

3.2 Day of maximum and minimum

As it was discussed in the introduction, the APD of each day of the week from the weekly average was calculated from the $d_{ijmy,new}$ values for each grid cell separately. The days of maximum and minimum APD, for all the grid cells with available data, are presented in Fig. 3a and c for MODIS TERRA and Fig. 3b and d for MODIS AQUA. Grid cells with a statistically significant APD at the 90% confidence level according to the two tailed t-test appear with a thin white outline. The patterns in TERRA and AQUA case are quite similar. The patterns are not very smooth for the day of maximum APD. However, one can distinguish a midweek (Wednesday-Thursday-Friday) plume in the biggest part of CE with a small accumulation of weekend (Saturday-Sunday-Monday) grid cells on the north-eastern part of the region. SWE on the other hand presents a clear weekend maximum. NEE and CEE are mostly covered by weekend grid cells with only fewer neutral (Tuesday) and weekday grid cells. The same stands for CM and EM but here the weekday grid cells are more and the patterns appear noisier. The day of minimum APD patterns are clearer. A Monday minimum plume is situated over CE extending partly into CEE. A midweek minimum appears over SWE and NEE. For CM and EM the majority of the grid cells present a neutral or midweek minimum. The existence of a neutral zone extending in the east of CE is indicative of an easterly moving weekly plume due to the prevailing synoptic conditions as it is discussed below.

Spatially averaged APDs were calculated using Eq. (3), and the weekly variability for each region, for both TERRA and AQUA MODIS, was examined (Fig. 4). In

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the case of CE a clear Monday minimum APD appears for both TERRA and AQUA datasets ($-5.54\%/-6.30\%$). Saturdays and Sundays present a low positive or near zero APD. The variability between midweek days ($\sim 1\%$) is very small compared to the difference between the day of minimum and maximum in both cases ($8.04\%/9.11\%$).
5 On the contrary, SWE exhibits a minimum APD on Thursday ($-3.89\%/-5.89\%$) and a maximum on Sunday ($+5.42\%/+4.75\%$). NEE exhibits a minimum APD on Friday (-4.85%) and a maximum on Sunday ($+8.05\%$) for TERRA MODIS. For AQUA, the day of minimum is Friday (-6.53%) and the day of maximum is Tuesday ($+3.68\%$) with a second maximum of $+3.47\%$ on Sunday. For CM, EM and CEE the weekly variability
10 is generally negative but it is also much less stronger than it is for the previous three regions. Among the last three regions, CEE presents the strongest weekly variability with a minimum APD during Tuesday ($-1.38\%/-2.83\%$) and a maximum during Sunday ($+3.56\%/+4.17\%$). The day of maximum and minimum APD along with the maximum and minimum APD values with the statistical significance at the 90% confidence level
15 and their corresponding differences are cited in Table 2.

3.3 A second approach for the investigation of weekly cycle

As it was discussed in Sect. 3.1., the region selected by Quaas et al. (2009) contains sub-regions with differing weekly cycles. It is shown here that the method could be better applied only after a detailed spatial analysis where regions with common characteristics have been specified. The spectral analysis applied by Quaas et al. (2009) indicated no statistical significance for the 7-day cycle of MODIS TERRA and AQUA AOD_{550} time series. However, as it is shown below, for smaller regions in Europe with a strong and uniform weekly cycle the spectral analysis indicates a statistically significant AOD_{550} 7-day cycle.
20

Daily area averages have been calculated for each one of the 6 box regions. 12 time series (6 for TERRA and 6 for AQUA) were then analyzed separately. The APD for the 7, 6 and 8-day weeks appears in Fig. 4. Each subfigure has been put next to the corresponding figure that emerged from the spatial analysis, for easier comparison of
25

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the two methods. The day of maximum and minimum percentage deviation along with the deviation values, their statistical significance at the 90% confidence level, the difference between maximum and minimum APD and the corresponding difference for the 6 and 8-day weeks are cited in Table 2 together with the corresponding results from the spatial analysis. The APDs emerging from the method presented in this paragraph are in general in good agreement with the results from the spatial analysis. Each method has its own advantages and disadvantages. Indicatively we could say that the spatial analysis captures the relative contribution of each grid cell to the APD in a better way. On the other hand, the second approach could eliminate the contribution of episodically high values which could insert drawbacks because the data are spatially averaged before the calculation of the APD. A great advantage of the second approach is that it allows for a further investigation of the statistical significance through the spectral analysis technique (Sect. 2.2) which was applied on the time series. Whether the 7-day period is statistically significant at the 90% or 95% confidence level or not is indicated in Table 2.

For CE, a clear midweek peak with a Monday APD minimum appears for both TERRA and AQUA datasets ($-4.34\%/-4.30\%$). Saturdays and Sundays present a low positive or near zero APD for TERRA MODIS. AQUA MODIS Sundays, in contrast to the spatial analysis, present a highly negative APD. Thursday is the day of maximum APD ($+2.08/+3.27\%$) in both datasets. The difference between the day of minimum and maximum is higher in the case of AQUA (7.57%) than TERRA (6.42%) which is in agreement with the spatial analysis. All the minima and maxima are statistically significant at the 90% confidence level. The statistical significance is supported by the fact that for both TERRA and AQUA the 7-day signal is much stronger than the 6 and 8-day signal as it is shown in Fig. 4b and d. The spectral analysis has shown that the 7-day cycle is statistically significant at the 95% confidence level for both datasets. In agreement to the weekly variability from the spatial analysis, SWE presents a clear weekend peak with minimum APD on Thursday ($-4.07\%/-5.67\%$) and a maximum on Sunday ($+4.32\%/+4.32\%$). The minima and maxima are statistically significant at the 90%

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minimum and weekend maximum is preserved. Moreover, the variability is rather different for the TERRA and AQUA MODIS datasets (Fig. 4v and 4x). This, in addition to the fact that the 7-day signal is stronger than 6 and 8-day signal in the case of TERRA and weaker in the case of AQUA MODIS does not allow for clear conclusions about the weekly variability over this region. The 7-day cycle is not statistically significant according to the spectral analysis. It has to be mentioned here that the significance of a 6 and 8-day cycle was investigated through the spectral analysis for the total of the time series and in every single case statistical significance was not indicated.

3.4 Seasonal variability

10 Apart from the identification of the general weekly cycle patterns and the investigation of the weekly variability of AOD_{550} over Europe the same quantities should also be examined on a seasonal basis. In this way, one can discriminate whether the weekly patterns are stable throughout the year and connect spatial and temporal variations with the dominating synoptic conditions in the greater area. The seasonal WCI patterns for MODIS TERRA and AQUA appear in Fig. 5. The average seasonal WCI values for the 6 box regions along with the seasonal AOD_{550} and FMR_{550} levels are cited in Table 3. The $1 \times 1^\circ$ MODIS AOD_{550} dataset has been shown to give a good insight into the regional aerosol levels for Europe (Chubarova, 2009). A comparison of the seasonal WCI maps (Fig. 5) with the seasonal AOD_{550} maps (not shown here), shows that strong positive/negative WCI patterns do not correlate with the high/low AOD_{550} patterns. Even for highly populated megacities (e.g. Moscow) the WCI does not reveal a strong weekly variability. Compared to the general WCI patterns, the seasonal WCI patterns appear noisier for DJF, MAM and SON. This should partially be expected since the seasonal averages are calculated from a fraction of the total WCI values and though more vulnerable to episodic events (e.g. dust outbreaks, forest fires, storms, etc.) which could lead to occasional high/low values or individual “bad” AOD_{550} retrievals which could lead to fault WCI values. The WCI patterns are noisier for the northern and north-eastern parts of Europe which are the regions with typically fewer WCI values. On the

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contrary, the summer WCI patterns are very clear and strong, driving the general WCI patterns, as it is shown below. The APD for the 6 box regions, as this was calculated from the TERRA and AQUA time series, appears in Fig. 6. The day with maximum and minimum APD and the corresponding APD values are cited in Table 3.

In this work, we present an effort to correlate differing weekly cycle patterns appearing over Europe with the dominating seasonal wind patterns. We suggest that aerosol weekly cycles could be transferred away from areas with strong and dominating weekly patterns. For areas away from anthropogenic activities or urban/sub-urban regions with a weak aerosol weekly variability, synoptic transport could be critical as far as weekly cycle patterns are concerned. For this reason, the wind speed and direction for the 850 mbar pressure level from the 2000–2009 NCEP/NCAR reanalysis data set was used. The average seasonal wind vectors at the 850 mbar pressure level were calculated (Fig. 7a, c, e, g) together with the frequency distribution of the wind direction over the 6 box regions (Fig. 7b, d, f, h). A detailed discussion for the seasonal weekly cycle patterns is presented below.

3.4.1 Winter

No conclusions can be reached for northern Europe during DJF, since MODIS aerosol retrievals are not feasible during winter for a large part of this area (see Fig. 5a and b), due to dominating cloud or snow/ice coverage (Remer et al., 2005). In addition, at low aerosol loading (optical thickness less than 0.15) which is the case for winter in this area, because of less signal, there is greater susceptibility to all algorithmic and sensor uncertainties (Remer et al., 2005).

Although during winter the aerosol patterns over CE appear rather noisy, there still seem to be areas in the center of this region with a strong positive cycle. The average WCI for DJF is +2.82% for TERRA and +3.35% for AQUA. There is a positive WCI plume extending over CEE (+7.31%/+15.09%) and parts of EM (+0.83%/+2.99%) (Figs. 4a and 4b). The fact that the highly positive and statistically significant grid cells are situated strictly over land areas over the whole Balkan Peninsula and the region of

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safe conclusion cannot be reached because as it was aforementioned the winter patterns are rather noisy for both CE and SWE. The weekly variability revealed through the APDs appearing in Fig. 6a–l is quite different for MODIS TERRA and AQUA time series in many cases. The variability is not clear for CE, Monday being the day of minimum APD for MODIS TERRA. In the case of AQUA the weekly variability is clear, with a Friday–Saturday peak, Monday being the day of minimum again. The corresponding APD quantities for the total of the regions are cited in Table 3. A clear weekly variability similar for TERRA MODIS and AQUA appears only in the case of EM. The variability is characterized by a Friday maximum and a Tuesday minimum for both datasets and is driven by the plume extending from the Balkans to the Libyan Sea.

3.4.2 Spring

During spring, a positive WCI plume appears in the center of CE with a strong negative cycle in the north-eastern part of this area (Fig. 5c and d). The strong positive plume covering regions situated over the region defined by the triangle of France–Germany–Switzerland is responsible for the positive mean WCI levels of CE (+1.60%/+2.71%). Winds are much weaker than during DJF (see Fig. 7a and c) which favours the development of a local weekly cycle in a region with or near regions with a strong industrial activity and/or high population. The weekly patterns are rather noisy over SWE. The average WCI is -1.96% and -1.41% for MODIS TERRA and AQUA. The same stands for NEE and CEE with different results for MODIS TERRA and AQUA. The weekly cycle is negative for MODIS TERRA and positive for MODIS AQUA for both regions (see Table 3). The APD variability does not give a clear variability for NEE. On the other hand, CEE presents a clear variability for both datasets with weekday minima and weekend maxima (Fig. 6i and j) A negative weekly cycle plume is observed over CM giving average WCIs of -4.48% (TERRA) and -6.07% (AQUA). The average WCIs are also negative for EM for both the instruments ($-4.00\%/-2.96\%$) with a strong negative plume covering the region defined by southern Italy, Adriatic Sea and southern Balkans. A weaker positive plume appears over the Libyan Sea primarily for the MODIS AQUA

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dataset. The negative plume appearing over CM and EM could be explained through the synoptic wind conditions which favour the transport of aerosols from CE westwards (see Fig. 7c and d). During spring, the wind direction ranges from west (W) to north-east (270° – 45°) with an average speed of ~ 7.8 m/s for $\sim 40\%$ of the days for CE. The corresponding wind speed and frequency values for CM are ~ 7 m/s and $\sim 53\%$. Following the same logic as in Sect. 3.4.1., the air masses arriving over CM from CE are expected to travel ~ 1 – 2 days on average. This shift is in good agreement with the weekly variability shown in Fig. 6a, b, g, h, taking into account that a weekly variability of aerosols is not expected over CM. The day of maximum for CE is Saturday while the days with maximum APDs in CM are Sunday and Monday. For EM, the wind direction ranges from west to north (270° – 360°) directions with an average speed of ~ 6.3 m/s for $\sim 40\%$ of the days. The estimated average time for the transport of air masses from CE is ~ 2 – 3 days. This is in good agreement with the 3 days difference appearing in the day of minimum APD for EM and CE (Thursday/Monday). In order to verify this, we applied a lag-correlation on the CE and EM time series after deseasonalizing them by removing the monthly mean values. The correlation coefficient calculated assuming a lag of 3 days between the time series is higher than the coefficients calculated for other lags from 0 to 20 (Fig. not shown here). The continental character of the two negative WCI plumes is verified from the fact that the plumes are still visible on the maps (see Fig. 8c and d) when applying the aerosol flags.

3.4.3 Summer

In contrast to the noisy patterns appearing in winter and spring, the summer WCI patterns are very clear. A comparison of the summer (Fig. 5e and 1f) and the general WCI patterns (Fig. 1c and d), shows that the summer weekly variability determines the annual weekly cycle. A very clear and distinct positive weekly cycle plume covers the whole CE extending over the British Islands to the north and over CM to the south. The WCI values are high ($+6.79\% / +6.77\%$) and the strong weekly cycle is reflected to the clear and strong weekly variability shown in Fig. 6a and b. A Thursday maximum and

a Sunday–Monday minimum with a max/min difference of 8–9% is observed for both MODIS TERRA and AQUA datasets. The extension of the Central-European plume to the south, perfectly matches the strong north-western mean flow at the 850 mbar pressure level appearing in Fig. 7e. Over CE, wind direction ranges from south-west to north (225°–360°) for the 62% of the days with an average wind speed of ~ 7.1 m/s, bringing air masses from the CE to NEE, CEE, CM and EM (Fig. 7f). A strong negative WCI plume covers the whole NEE region extending to the northwest. The average WCI value is almost the same for TERRA and AQUA dataset ($-12.16\%/-12.19\%$). The weekly variability shows a clear weekday minimum (Wednesday–Thursday–Friday) and a weekend maximum (Saturday–Sunday) with the max/min difference being in the order of ~ 13 –16% for the two instruments (Table 3). ~ 2 –3 days is the expected time of the air mass transport from CE to NEE, which is supported from the results of the lag-correlation applied on the time series. The winds are relatively low over NEE, favouring the development of a weekly cycle. As it is apparent from the WCI levels over CEE, winds originating from CE could be responsible for the negative cycle appearing there. A Friday minimum and a Sunday maximum appears in both MODIS TERRA and AQUA datasets. During late summer the existence of a significant number of agricultural fires in the region around Ukraine is typical and could possibly impact the weekly variability. Like in the case of NEE, the weekly variability could be explained through the transport of air masses from the westerly dominating wind flow. The ~ 2 –3 days difference between the CE and CEE maxima is also supported by a 2–3 days lag-correlation maximum of the time series (figure not shown here). A very interesting feature of Fig. 5e and f is the strong negative WCI plume ($-11.03\%/-9.62\%$) covering the region around the Iberian Peninsula. This feature would not be expected to originate from CE since winds with northern to eastern direction dominate only for the ~ 12 –13% of the days in the region of CE and CM. Winds blowing from the opposite direction account for only $\sim 8\%$ of the days, which rejects the assumption that aerosols could be transferred from the Algerian coast. Moreover, the maps with the WCI patterns after the application of the aerosol flags (see Fig. 8e and f) reveal that the WCI patterns still remain negative.

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This shows that the negative weekly cycle is possibly of continental origin. The fact that the statistical highly negative grid cells are located over land and that winds are relatively mild over the region favours the assumption that this is a local weekly cycle. The weekly variability over EM is very low during summer. The APD values appearing in Fig. 6i and j are quite similar for TERRA and AQUA but the weekly variability is very weak. On the contrary the weekly variability over CM is strong with minimum APDs appearing on Saturday and Sunday and maximum APDs on Monday and Tuesday. CM seems to be influenced from air masses from both SWE and CE which lead to two separate plumes in the region. The north-eastern part is covered by a positive WCI plume while the south-western part from a strong negative WCI plume. Taking into account that over CM the wind direction ranges from west to north (270° – 360°) for ~48% of the days with an average speed of ~ 5.6 m/s during summer, the air masses originating from SWE and CE are supposed to arrive over CM within a day. The negative plume is probably the extension of the plume appearing over SWE and the positive an extension of the CE plume. The weekly variability shown in Fig. 6g and h is probably the result of the synoptic aerosol transport from SWE and CE with a maximum APD on Monday-Tuesday and a minimum on Saturday-Sunday.

3.4.4 Autumn

During autumn, the WCI patterns (Fig. 5g and h) share characteristics with the summer and spring WCI patterns. A significant positive WCI plume appears over CE (+4.02%/+7.60%). The weekly variability is clear, being characterized by a mid-week maximum (Wednesday-Thursday) and a weekend minimum (Saturday-Sunday-Monday) (Fig. 6a and b). The weekly patterns are negative over SWE with the average WCI being -4.63% and -4.21 for MODIS TERRA and AQUA. The weekly variability here is very close to the summer variability (Fig. 6c and d), with a midweek minimum on Thursday and a weekend maximum on Monday with a max/min difference of $\sim 10\%$ and $\sim 15\%$ for TERRA and AQUA data correspondingly. As it is revealed from the WCI patterns and the weekly variability (Fig. 5g, h, 6e, f, k, l) the weekly cycle is not very

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clear over NEE and CEE. However, the average WCI values are negative for both the datasets which is in agreement to the summer values. The significant negative plume appearing over EM could be explained through the synoptic wind conditions which favour the transport of aerosols from CE westwards (Fig. 7g and h). During autumn, the wind direction ranges from west to north (270° – 360°) for ~30% of the days with relatively an average speed of ~9 m/s for CE (Fig. 7h). The corresponding values for EM are ~37% and ~5.6 m/s. The result is a large significant negative weekly cycle appearing over the Balkan Peninsula and the Libyan Sea. This plume determines the weekly variability over EM. The lag-correlation between the CE and EM time series reveals a peak for a 3–4 days lag (figure not shown here). This is in good agreement with the reverse weekly variability appearing for EM and CE (Fig. 6a, b, i, j) and could explain the negative plume appearing over EM through the transport of continental aerosols from CE. A further indication of the continental origin of the plume is its persistence after the subtraction of the days where dust aerosols were the dominant type (Fig. 8g and h). A strong positive WCI plume appears over CM. The patterns are similar for MODIS TERRA and AQUA with the average WCI levels being +2.41% and +3.27% respectively over CM. This plume is possibly an extension of the CE plume and its continental origin is verified by its persistence after the application of the aerosol flags (Fig. 8g and h).

20 4 Summary and conclusions

In the present study, a satellite-based index (WCI) expressing the mean percentage of the AOD_{550} difference between midweek and weekend to the weekly average was used for the spatial and temporal investigation of the regional patterns of aerosol weekly cycle over Europe. This method was originally used by Xia et al. (2008); however 25 it is shown here that the method has limitations which should be taken into account when being applied. In order to remove episodic dust transport events, 2 different aerosol flags, employing FMR_{550} and AOD_{550} data, were applied diagnostically. After

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Table 1. Regions of interest defined from the WCI with their corresponding geolocation, the average WCI per region for the initial dataset, for the corrected dataset using the $\pm 1000\%$ limit and for the corrected dataset using the $\pm 1000\%$ limit and additional corrections using flag1 and flag2. The statistical significance at the 90% confidence level and the number of values included in the calculations are also given.

Region	Geolocation	(TERRA) WCI/WCI corr. (%) WCI corr. & flag1/2	(TERRA) Signif. 90%	(TERRA) Values	(AQUA) WCI/WCI corr. (%) WCI corr. & flag1/2	(AQUA) Signif. 90%	(AQUA) Values
Q90 EUROPE (OE)	(35° N–70° N, 10° W–30° E)	-0.80/-0.58 +0.33/+0.21	Yes/yes Yes/yes	391 469/391 352 354 852/359 811	-0.23/-0.07 +0.53/+0.40	No/no Yes/yes	277 522/277 418 255 849/259 122
Central Europe (CE)	(45° N–55° N, 3° W–23° E)	+4.28/+4.13 +4.06/+4.06	Yes/yes Yes/yes	65 158/65 146 57 027/58 181	+5.56/+5.60 +5.11/+5.20	Yes/yes Yes/yes	44 693/44 681 40 187/40 968
South-western Europe (SWE)	(36° N–44° N, 11° W–0° E)	-5.58/-5.71 -3.53/-3.55	Yes/yes Yes/yes	36 169/36 158 33 162/33 394	-4.81/-3.87 -1.19/-1.45	Yes/yes Yes/yes	25 827/25 802 23 925/24 066
North-eastern Europe (NEE)	(51° N–58° N, 25° E–50° E)	-9.17/-6.98 -6.68/-6.87	Yes/yes Yes/yes	34 663/34 625 32 636/32 893	-6.56/-6.88 -6.88/-6.86	Yes/yes Yes/yes	23 730/23 697 22 799/22 934
Central Mediterranean (CM)	(35° N–44° N, 2° E–15° E)	-1.28/-1.28 +0.57/+0.45	Yes/yes Yes/yes	50 386/50 386 43 711/44 414	-0.48/-0.53 +0.34/+0.28	No/yes No/yes	36 191/36 190 31 771/31 195
Eastern Mediterranean (EM)	(31° N–43° N, 17° E–37° E)	-1.87/-1.86 +0.06/-0.06	Yes/yes No/no	99 464/99 463 85 832/87 187	-1.79/-1.82 +0.61/+0.26	Yes/yes Yes/no	71 526/71 525 62 539/63 473
Central-Eastern Europe (CEE)	(44° N–50° N, 25° E–50° E)	-1.20/-1.10 +0.02/+0.02	Yes/yes No/no	45 716/45 699 42 347/42 756	-1.00/-0.47 +0.68/+0.40	No/no No/no	31 438/31 405 29 310/29 591

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Table 2. Day of maximum and minimum average percent departure from the weekly mean calculated from the spatial analysis (APD) and the corresponding APD values, day of maximum and minimum average percent departure from the weekly mean calculated from the region time series (APD_{ts}) and the corresponding APD_{ts} values, statistical significance at the 90% confidence level and the number of values included in the calculations, maximum and minimum APD calculated from the time series assuming hypothetical 6 and 8-day weeks, statistical significance of a 7-day cycle (90%/95% confidence level) indicated from the spectral analysis applied on the time series.

Region	(TERRA) Max/min day (APD%) Max/min day (APD _{ts} %)	(TERRA) Max-min APD diff. Max-min APD _{ts} diff.	(TERRA) Signif. 90% Signif. 90% 6/8d. signal Sp. analysis	(TERRA) Values	(AQUA) Max/min day (APD%) Max/min day (APD _{ts} %)	(AQUA) Max-min APD diff. Max-min APD _{ts} diff.	(AQUA) Signif. 90% Signif. 90% 6/8d. signal Sp. analysis	(AQUA) Values
CE	Wed(+2.50)/Mon(−5.54) Thu(+2.08)/Mon(−4.34)	8.04 6.42	Yes/yes Yes/yes (2.52/2.15) 95%	42 995/41 702 458/452	Fri(+2.81)/Mon(−6.30) Thu(+3.27)/Mon(−4.30)	9.11 7.57	Yes/yes Yes/yes (1.44/4.07) 95%	29 444/ 29245 336/334
SWE	Sun(+5.42)/Thu(−3.89) Sun(+4.32)/Thu(−4.07)	9.31 8.39	Yes/yes Yes/yes (4.12/7.70) 95%	26 627/27 112 458/459	Sun(+4.75)/Thu(−5.89) Sun(+4.32)/Thu(−5.67)	10.64 9.99	Yes/yes Yes/yes (4.71/6.60) —	18 959/ 19415 335/337
NEE	Sun(+8.05)/Fri(−4.85) Sun(+3.74)/Fri(−3.10)	12.90 6.84	Yes/yes No/no (4.82/7.96) —	22 527/ 24 301 320/324	Tue(+3.68)/Fri(−6.53) Sun(+5.00)/Fri(−4.94)	10.21 9.94	Yes/yes Yes/yes (7.50/10.25) —	15 151/ 16230 219/226
CM	Sun(+2.47)/Tue(−1.33) Sun(+1.92)/Tue(−1.35)	3.80 3.27	Yes/yes No/no (1.49/6.48) —	37 652/ 36 950 457/456	Mon(+1.37)/Fri(−1.43) Sat(+1.07)/Fri(−1.51)	2.80 2.58	Yes/yes No/no (2.14/7.73) —	26 709/26 859 337/337
EM	Sun(+2.03)/Wed(−1.57) Sun(+2.50)/Wed(−1.89)	3.60 4.39	Yes/yes Yes/yes (2.10/7.09) 95%	76 527/77 363 458/459	Sun(+3.04)/Tue(−2.76) Sun(+2.58)/Tue(−2.89)	5.80 5.47	Yes/yes No/no (2.25/6.60) 90%	54 229/52 819 335/336
CEE	Sun(+3.56)/Tue(−1.38) Sun(+3.19)/Sat(−2.55)	4.94 5.74	Yes/yes Yes/yes (3.36/4.86) —	32 517/31 716 451/454	Sun(+4.17)/Tue(−2.83) Sun(+1.43)/Thu(−1.61)	7.00 3.04	Yes/yes No/no (5.16/6.66) —	22 148/21 566 334/335

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Table 3. Average seasonal WCI per region for the $\pm 1000\%$ limit corrected dataset with the average AOD_{550} and FMR_{550} levels, seasonal day of maximum and minimum average percent departure from the weekly mean calculated from the region time series (APD_{ts}) and the corresponding APD_{ts} values, statistical significance at the 90% confidence level and the number of values included in the calculations.

Region/ Season	(TERRA) WCI (%) (AOD/FMR)	(TERRA) Signif. 90%	(TERRA) Values	(TERRA) Max/min day (APD _{ts} %)	(AQUA) WCI (%) (AOD/FMR)	(AQUA) Signif. 90%	(AQUA) Values	(AQUA) Max/min day (APD _{ts} %)
CE/DJF	+2.82 (0.15/0.40)	Yes	6070	Thu(+1.99)/Mon(−4.81)	+3.35 (0.13/0.47)	Yes	2561	Sat(+5.56)/Mon(−5.18)
CE/MAM	+1.60 (0.26/0.57)	Yes	19 539	Sat(+3.40)/Mon(−6.94)	+2.71 (0.25/0.62)	Yes	12 798	Sat(+5.82)/Mon(−8.54)
CE/JJA	+6.79 (0.26/0.55)	Yes	22 345	Thu(+4.73)/Mon(−3.38)	+6.77 (0.25/0.62)	Yes	17 249	Thu(+4.82)/Sun(−4.01)
CE/SON	+4.02 (0.16/0.52)	Yes	17 169	Wed(+4.35)/Mon(−2.30)	+7.60 (0.15/0.58)	Yes	12 003	Thu(+5.13)/Sun(−4.48)
SWE/DJF	−4.69 (0.11/0.39)	Yes	8696	Sun(+5.91)/Wed(−4.62)	+1.46 (0.10/0.44)	No	5731	Sun(+7.48)/Tue(−3.89)
SWE/MAM	−1.96 (0.20/0.36)	Yes	8516	Sun(+4.14)/Tue(−5.32)	−1.41 (0.19/0.39)	No	5785	Sun(+7.04)/Tue(−8.46)
SWE/JJA	−11.03 (0.21/0.31)	Yes	9588	Mon(+7.36)/Thu(−7.71)	−9.62 (0.20/0.34)	Yes	7356	Mon(+5.67)/Thu(−7.87)
SWE/SON	−4.63 (0.15/0.37)	Yes	9358	Mon(+4.87)/Thu(−5.32)	−4.21 (0.13/0.41)	Yes	6930	Mon(+8.07)/Thu(−6.60)
NEE/DJF	+1.84 (0.16/0.14)	Yes	43	Tue(+11.59)/Fri(−19.91)	−0.16(0.26)	—	0	—
NEE/MAM	−2.71 (0.22/0.65)	Yes	10 158	Sun(+2.40)/Tue(−3.22)	+1.30 (0.21/0.70)	No	5920	Wed(+3.96)/Fri(−2.59)
NEE/JJA	−12.16 (0.19/0.74)	Yes	15 835	Sun(+7.78)/Wed(−5.54)	−12.19 (0.19/0.84)	Yes	12 329	Sun(+10.69)/Thu(−4.97)
NEE/SON	−2.48 (0.16/0.52)	Yes	8563	Thu(+3.60)/Sat(−4.37)	−3.73 (0.16/0.61)	Yes	5432	Mon(+5.73)/Fri(−7.77)
CMD/JF	−3.29 (0.14/0.47)	Yes	12 451	Sat(+5.43)/Tue(−5.39)	−2.10 (0.13/0.48)	Yes	8377	Sat(+6.54)/Tue(−5.21)
CMM/MAM	−4.48 (0.26/0.49)	Yes	12 243	Fri(+4.43)/Wed(−6.89)	−6.07 (0.24/0.51)	Yes	8292	Sun(+5.26)/Wed(5.73)
CM/JJA	+0.02 (0.27/46)	No	12 809	Mon(+6.96)/Sat(−5.46)	+1.72 (0.26/0.47)	Yes	9710	Tue(+5.25)/Sat(−6.05)
CM/SON	+2.41 (0.20/0.51)	Yes	12 883	Fri(+1.56)/Mon(−3.84)	+3.27 (0.18/0.52)	Yes	9811	Sat(+1.88)/Sun(−3.53)
EM/DJF	+0.83 (0.18/0.49)	Yes	22 356	Fri(+4.44)/Tue(−5.55)	+2.99 (0.18/0.50)	Yes	15 186	Fri(+6.86)/Tue(−7.39)
EM/MAM	−4.00 (0.30/0.44)	Yes	24 714	Sun(+3.65)/Thu(−4.46)	−2.96 (0.29/0.46)	Yes	16 563	Sun(+2.24)/Thu(−3.48)
EM/JJA	+0.58 (0.26/0.43)	Yes	26 194	Fri(+1.88)/Mon(−1.28)	−0.67 (0.25/0.44)	Yes	19 932	Fri(+1.75)/Tue(−1.52)
EM/SON	−4.57 (0.21/0.48)	Yes	26 198	Sat(+2.82)/Fri(−3.14)	−5.70 (0.20/0.49)	Yes	19 842	Sat(+3.98)/Wed(−3.11)
CEE/DJF	+7.31 (0.14/0.57)	Yes	3992	Fri(+9.70)/Sat(−5.26)	+15.09 (0.13/0.59)	Yes	2217	Tue(+3.78)/Sat(−3.01)
CEE/MAM	−2.47 (0.21/0.69)	Yes	13 114	Sun(+6.21)/Sat(−6.07)	+0.74 (0.20/0.69)	No	8151	Sun(+3.98)/Thu(−2.46)
CEE/JJA	−1.47 (0.24/0.52)	Yes	15 826	Sun(+2.21)/Fri(−3.38)	−3.25 (0.23/0.58)	Yes	11 940	Sun(+2.55)/Fri(−2.49)
CEE/SON	−1.74 (0.16/0.54)	Yes	12 760	Mon(+3.09)/Wed(−3.88)	−2.22 (0.15/0.54)	Yes	9053	Sat(+6.33)/Wed(−4.23)

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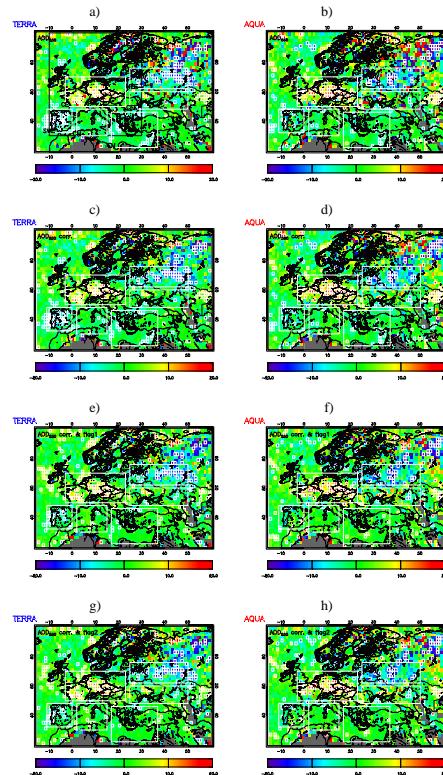


Fig. 1. Left column: From TERRA MODIS data for the period February 2000–February 2009. Right column: AQUA MODIS data for the period July 2002–December 2008. **(a, b)** WCI patterns calculated from AOD_{550} measurements, **(c, d)** the same as **(a, b)** but for data corrected with the $\pm 1000\%$ limit, **(e, f)** the same as **(c, d)** but for data corrected with flag1, **(g, h)** the same as **(c, d)** but for data corrected with flag2. The grid cells with a statistically significant value at the 90% confidence level according to the two tailed t-test appear with a thin white outline. The six regions of interest are marked with a thick white outline in all the maps. The region examined in Quaas et al. (2009) (QE) is marked with a thick black line in **(a)** only. See also text.

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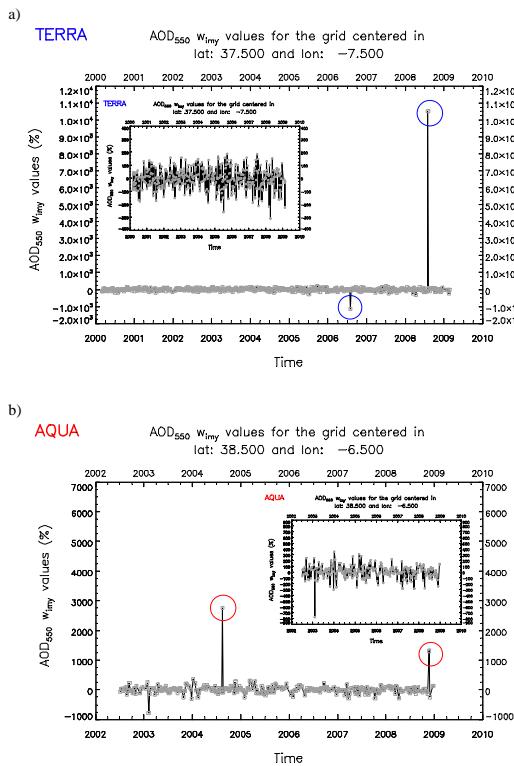


Fig. 2. w_{imy} time series before (big panel) and after (small panel) the application of the $\pm 1000\%$ limit correction for (a) TERRA MODIS AOD₅₅₀ measurements for the grid cell centered in (37.5° N, 7.5° W), (b) AQUA MODIS AOD₅₅₀ measurements for the grid cell centered in (38.5° N, 6.5° W). In both figures the values that exceed the $\pm 1000\%$ limit are encircled.

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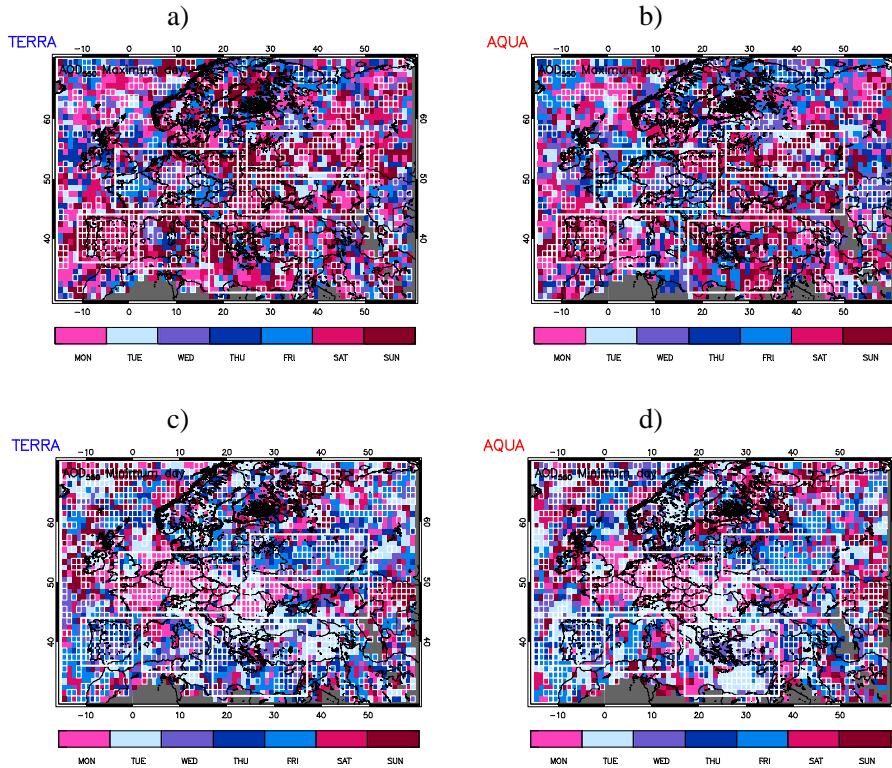


Fig. 3. (a, b) Day of maximum APD calculated from TERRA and AQUA MODIS AOD_{550} measurements, **(c, d)** Day of minimum APD calculated from TERRA and AQUA MODIS AOD_{550} measurements. The grid cells with a statistically significant value at the 90% confidence level according to the two tailed t-test appear with a thin white outline. The six regions of interest are marked with a thick white outline in all the maps.

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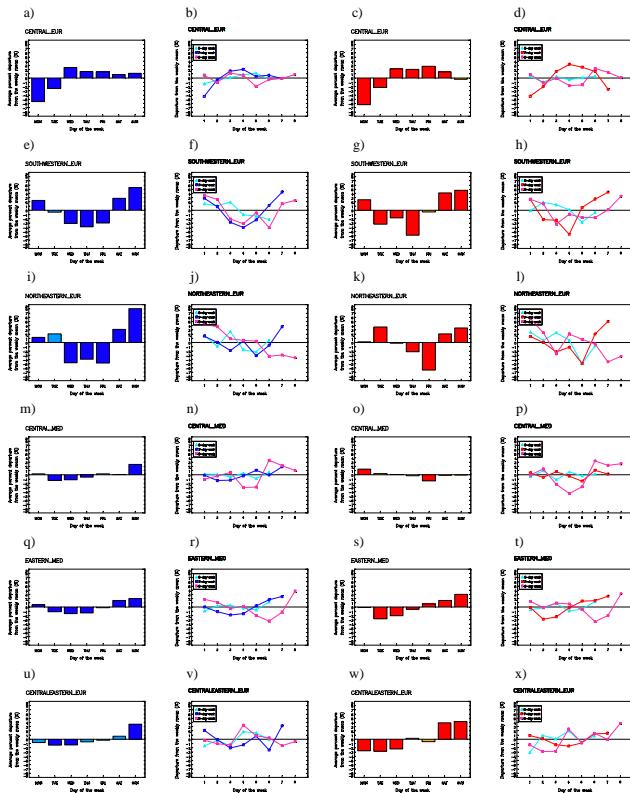


Fig. 4. Weekly variability of (a) spatially averaged APDs calculated using TERRA MODIS AOD₅₅₀ grid measurements for the period February 2000–February 2009 for CE, (b) APDs calculated from the CE spatially averaged time series for 7, 6 and 8-day weeks (TERRA MODIS), (c) the same as a but for AQUA MODIS measurements for the period July 2002–December 2008, (d) the same as b but for AQUA MODIS measurements for the period July 2002–December 2008, (e, f, g, h) the same as a, b, c, d but for SWE, (i, j, k, l) the same as a, b, c, d but for NEE, (m, n, o, p) the same as a, b, c, d but for CM, (q, r, s, t) the same as a, b, c, d but for EM, (u, v, w, x) the same as a, b, c, d but for CEE.

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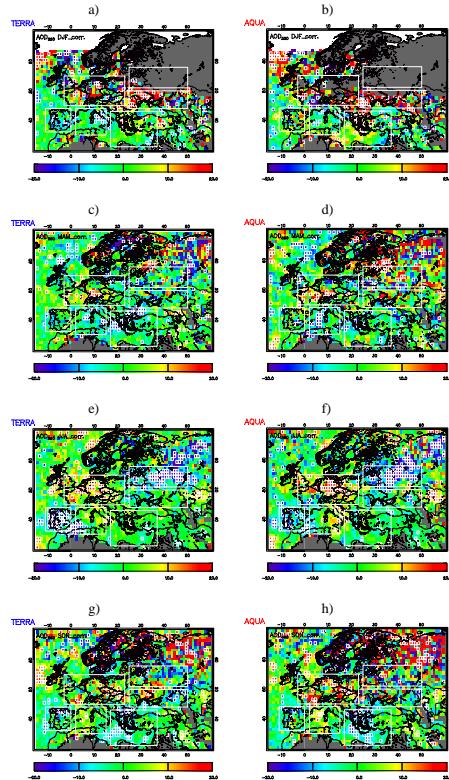


Fig. 5. Seasonal WCI patterns calculated from MODIS AOD₅₅₀ measurements corrected with the $\pm 1000\%$ limit (a) DJF TERRA (February 2000–February 2009), (a) DJF AQUA (July 2002–December 2008), (c) the same as a but for MAM, (d) the same as b but for MAM, (e) the same as a but for JJA, (f) the same as b but for JJA, (g) the same as a but for SON, (h) the same as (b) but for SON. The grid cells with a statistically significant value at the 90% confidence level according to the two tailed t-test appear with a thin white outline. The 6 regions of interest are marked with a thick white outline in all the maps.

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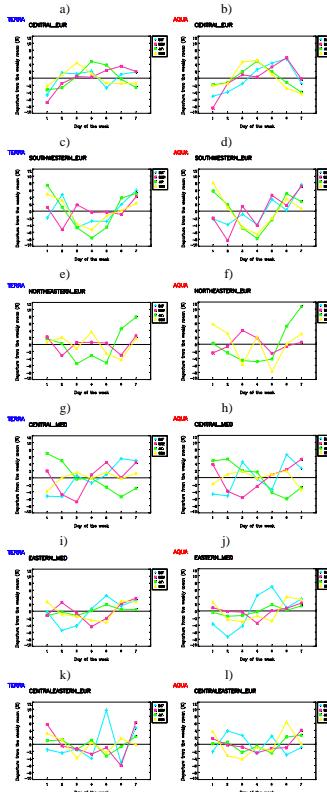


Fig. 6. Seasonal weekly variability of (a) APDs calculated from the CE spatially averaged time series for 7, 6 and 8-day weeks (TERRA MODIS: February 2000–February 2009), (b) APDs calculated from the CE spatially averaged time series for 7, 6 and 8-day weeks (AQUA MODIS: July 2002–December 2008), (c) the same as a but for SWE, (d) the same as b but for SWE, (e) the same as a but for NEE*, (f) the same as b but for NEE*, (g) the same as a but for CM, (h) the same as b but for CM, (i) the same as a but for EM, (j) the same as b but for EM, (k) the same as a but for CEE, (l) the same as (b) but for CEE. *Winter APD variability for NEE is not presented here due to the limited amount of AOD_{550} measurements.

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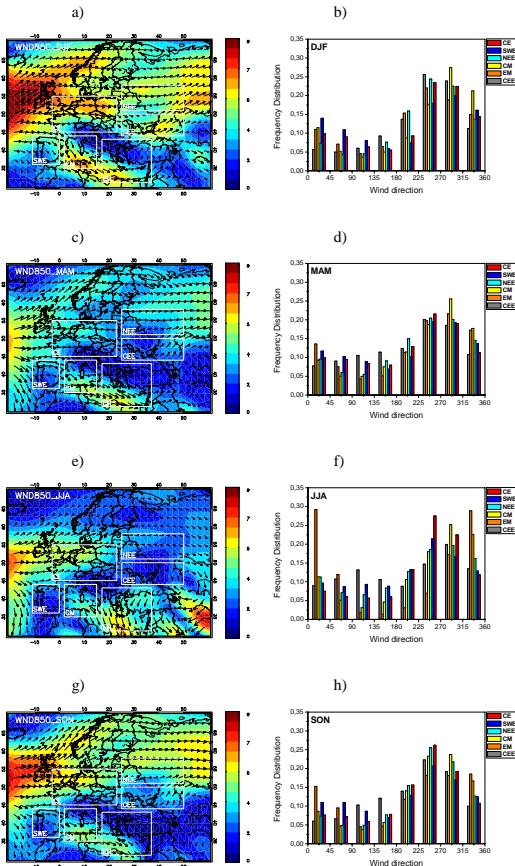


Fig. 7. (a) Average 2000–2009 NCEP/NCAR reanalysis wind speed (colorscale in m/s) and vectors at the 850 mbar pressure level for DJF, (b) Frequency distribution of the wind direction over the 6 regions of interest for DJF, (c) the same as a but for MAM, (d) the same as b but for MAM, (e) the same as a but for JJA, (f) the same as b but for JJA, (g) the same as a but for SON, (h) the same as b but for SON.

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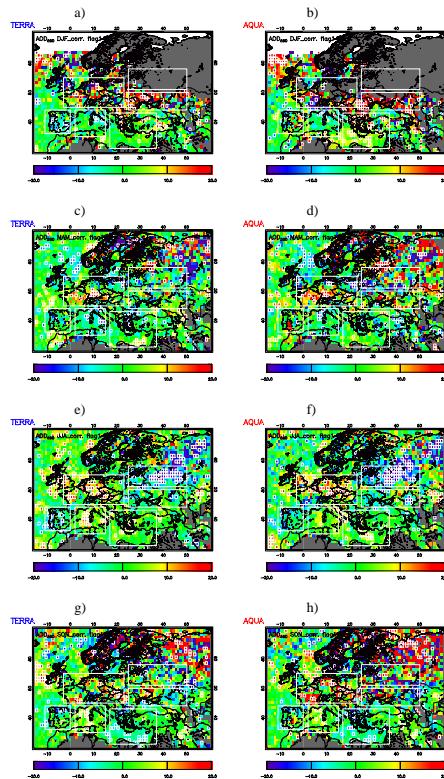


Fig. 8. Seasonal WCI patterns calculated from MODIS AOD₅₅₀ measurements corrected with the $\pm 1000\%$ limit and flag1 **(a)** DJF TERRA (February 2000–February 2009), **(b)** DJF AQUA (July 2002–December 2008), **(c)** the same as a but for MAM, **(d)** the same as **(b)** but for MAM, **(e)** the same as a but for JJA, **(f)** the same as **(b)** but for JJA, **(g)** the same as a but for SON, **(h)** the same as **b** but for SON. The grid cells with a statistically significant value at the 90% confidence level according to the two tailed t-test appear with a thin white outline. The 6 regions of interest are marked with a thick white outline in all the maps.