Mapping the physico-chemical properties of mineral dust in Western Africa: Mineralogical composition

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Abstract
In the last few years, several ground-based and airborne field campaigns have allowed exploring the properties and impacts of mineral dust in Western Africa, one of the major emission and transport areas worldwide. In this paper, we explore the synthesis of these observations to provide with a large-scale quantitative view of the mineralogical composition and its variability with source region and time after transport.

This work reveals that mineral dust in Western Africa is a mixture of clays, quartz, iron and titanium oxides, representing at least 92\% of the dust mass. Calcite ranged between 0.3 and 8.4\% of the dust mass depending on the origin. Our data do not show a systematic dependence of the dust mineralogical composition with origin, likely as in most of the cases they represent the composition of the atmospheric burden after 1–2 days after emission, when air masses mix and give raise to a more uniform dust load. This has implications for the representation of the mineral dust composition in regional and global circulation models, and satellite retrievals.

Iron oxides account for 58 ± 7\% of the mass of elemental Fe, and between 2 and 5\% of the dust mass. Most of them are composed of goethite, representing between 52 and 78\% of the iron oxide mass. We estimate that titanium oxides account for 1-2\% of the dust mass, depending on whether the dust is of Saharan or Sahelian origin.

The mineralogical composition is a critical parameter to estimate the radiative and biogeochemical impact of mineral dust. The results on dust composition have been applied to estimate the optical properties as so as the iron fractional solubility of Saharan and Sahelian dust.
Data presented in this paper are provided in numerical form upon email request while they are being implemented as a public database, the Dust-Mapped Archived Properties (DUST-MAP), an open repository for compositional data from other source regions in Africa and worldwide.

1. Introduction

Mineral dust from wind-driven soil erosion is an important player in the climate system. It is emitted in large quantities from arid and semi-arid regions of the globe, mostly Africa, Asia, Australia and North America (Shao et al., 2011), accounting for about 2000 Mt to the annual aerosol emissions at the global scale. Once in the atmosphere, mineral dust has several climatic and environmental impacts, related to its ability of scattering and absorbing radiation, both in the solar and terrestrial spectra, of acting as giant cloud and ice nuclei, of altering the concentrations of some gaseous pollutants (such as ozone), and of providing nutrients to the ecosystems via dry and wet deposition, whereby also changing the surface albedo (Shao et al., 2011; Mahowald et al., 2011). The relevance of those phenomena can be expressed in terms of radiative forcing, whose magnitude, in terms of central values, is estimated to be as high as 0.5–1 W m⁻² at the global scale (Mahowald et al., 2011).

However, uncertainties remain important, because of the spatial heterogeneity of concentrations, due to localized emission and short residence time in the atmosphere (1 week at the most, depending on particle size), but also because of the far from satisfactory knowledge on the underlying physico-chemical properties, composition, size and shape, ruling the optical, chemical and physical interactions of mineral dust with radiation and the atmospheric/terrestrial/oceanic constituents (Formenti et al., 2011a).

This is particularly the case of the composition. Dust is made of different minerals, whose proportions, at emission, depend on the mineralogy of the source region and surface wind speed relative to the soil roughness, which determines their size distribution (Marticorena and Bergametti, 1995, Alfaro et al. 1998). The various minerals, clays, quartz, potassic and sodic feldspars, calcium-rich carbonates and sulphates, iron and titanium oxides, have different mineralogical and crystallographic properties, therefore they may act differently with respect to light extinction, absorption of water, and surface reactivity (Pye, 1987). For example, iron oxides have been shown to control the interaction with light in the UV-visible spectrum (Sokolik and Toon, 1999), whereas quartz, clays and Ca-rich carbonates become important when looking at the infrared part of the radiation spectrum (Sokolik et al., 1998). Clays but also feldspars come to play when it comes in investigating the capacity of dust of activating as ice nuclei at low temperature (Hoose et al., 2008; Atkinson et al., 2013), whereas calcite (calcium carbonate) shows high surface reactivity with respect to various atmospheric pollutants (Crowley et al., 2010). Another clay family, the smectites, bears the
most soluble iron which might become available as nutrient to the marine phytoplankton (Journet et al., 2008). The properties of mineral dust are size-dependent, and the dimensional spectrum of mineral dust particles at emission extends over various orders of magnitude from approximately 200-300 nm to 100 µm (Kandler et al., 2009) depending on wind speed at emission. As a consequence, the properties of mineral dust vary as a function of time after emission due to size-dependent deposition, mixing, and gaseous condensation. Because these processes are not well established, the physico-chemical properties of mineral dust have to be determined experimentally at the various times of their life cycle.

In this paper, we relate on the mineralogical and chemical composition at emission or at most within 1–2 days from emission of mineral dust over Western Africa. The southern Sahara is the location of two of most persistent dust sources at the global scale, the Bodélé depression in between the Tibesti and the Ennedi mountain chains in Chad, and the areas in between the east of Mauritania, the North of Mali and the south of Algeria (Prospero et al., 2002; Laurent et al., 2008; Washington et al., 2006; Shao et al., 2011). These sources are relevant at the global scale as their emissions are transported across the Atlantic Ocean towards South and Central America (Reid et al., 2003; Koren et al., 2006; Ben-Ami et al., 2009; 2010).

The atmospheric content of mineral dust in Western Africa shows a very pronounced seasonal cycle despite high variability, both at the daily and inter-annual scales (N’Tchayi et al., 1994; N’Tchayi Mbourou et al., 1997; Goudie and Middleton, 2001). This is largely explained by the alternance of two meteorological regimes. Throughout the year, the “Harmattan”, a north-eastern dry wind, is responsible for intense dust emissions in the arid areas of the Sahara and very efficient transport towards the Atlantic Ocean and across the Sahel (Pye, 1987; Sokolik et al., 2001; Laurent et al., 2008). Wintertime is also the biomass burning season south of approximately 13°N (Haywood et al., 2008), and mixing between dust and biomass burning aerosols might occur during southward transport (Johnson et al., 2008). During summer, the Inter Tropical Convergence Zone (ITCZ) displaces northward, allowing the surface Monsoon flow to intrude the continent from the South-West towards the continental Sahel. This shift induces a progressive organization of convection from isolated cells to large-scale fast propagating systems, which, particularly at the beginning of the Monsoon season, are often not precipitating, and efficient in eroding bare soils and producing mineral dust (Marticorena et al., 2010; Abdourhamane Touré et al., 2011). The summertime column dust content over western Africa results therefore from the superimposition of local emission from the Sahel and remote transport of dust emitted in the Sahara (Tegen and Fung, 1994; Yoshioka et al., 2005; Marticorena et al., 2010).
The mineralogy of the Saharan and Sahelian source areas have different mineralogy: the Sahara is mainly composed by calcsols or arenosols, whereas the Sahel is richer in weathered plinthosols (Fe oxides/kaolinite/quartz) (Pye, 1987; Claquin et al., 1999; Caquineau et al., 2002; Nickovic et al., 2012; Journet et al., 2014). As a consequence, mixing or layering of dust transported from the Sahara and locally emitted by convection over the Sahel should be detectable through differences in the composition over the atmospheric column.

Furthermore, western Africa is prone to land-use change because of the rapid demographic increase and desertification during drought periods (ECA, 2005; Dai, 2011), which could feedback in increased emission (Carslaw et al., 2010). Rajot (2001) have shown that, in the Sahel, additional emission by wind erosion of mesoscale meteorological disturbances, such as organized convective systems, occurs on harvested soils only. These additional emissions, anthropogenic in origin, are not quantified to date (Boucher et al., 2013).

To provide with the quantitative estimate of the mineralogical composition of mineral dust at the regional scale, we report in this paper the synthesis of data from various aircraft and ground-based field campaigns which have taken place in western Africa in 2006 and 2007. These are the African Monsoon Multidisciplinary Analysis (AMMA) Special Observing Period-0 (SOP0; Rajot et al., 2008), augmented of the airborne Dust and Biomass-burning Experiment (DABEX; Haywood et al., 2008), and Special Observing Period-1 and -2 (SOP1-2; Reeves et al., 2010); the Dust Outflow and Deposition to the Ocean (DODO; McConnell et al., 2008) and the Geostationary Earth Radiation Budget Intercomparison of Longwave and Shortwave radiation (GERBILS; Haywood et al., 2011).

Some of the data discussed in this paper have been published previously (Formenti et al., 2008; 2011b; Rajot et al., 2008; Klaver et al., 2011). Here, they are revisited with a broader perspective discussion of the mineralogical variability at the regional scale of Western Africa and its implications for climate.

This paper discusses whether, when mapped at the scale of western Africa, the regional variability at emission deriving from the heterogeneity of the mineralogy of the parent soils is still relevant or whether it becomes irrelevant when dust is airborne. This is important for developing a model parameterisation for dust in regional and global climate models or for satellite retrieval algorithms.

2. Methods

Full details of the field campaigns operations, experimental procedures, collection and analytical protocols are described in a number of companion papers (Haywood et al., 2008;
2.1. Field campaigns

The African Monsoon Multidisciplinary Analysis (AMMA) was an international project to improve our knowledge and understanding of the West African monsoon (WAM), its variability on daily-to-interannual time scales, and its effects on the status of the atmosphere (Redelsperger et al., 2006). Because of its links to precipitation, wind speed, soil surface state, and drought at different time scales, mineral dust aerosols play a role in this comprehensive scientific framework.

The experimental strategy of the AMMA program was based on embedded multi-year, seasonal and intensive observation periods (Janicot et al., 2008; Lebel et al., 2010). The observations used in this paper were conducted during the intensive observations periods, called Special Observing Periods (SOPs). The wintertime SOP (SOP0), was dedicated to the investigation of mineral dust and biomass burning, and their mixing, took place from 13 January and 13 February 2006. The summertime observations (periods SOP1 and SOP2) started on 31 May 2006 and ended on 17 July 2006 and were dedicated to the fine description of the interactions between aerosols and convection.

The ground-based observations were conducted at the AMMA supersite of Banizoumbou (13.5°N; 2.6°E, 250 m above sea level), located at a remote location at about 60 km east from the capital of Niger, Niamey. This site has been operational since the early 1990s, when the first measurements of soil erosion were performed on a cultivated field and a fallow (Rajot, 2001). Since 1995, the site is also an Aerosol Robotic Network (AERONET) station measuring columnar aerosol optical properties. The ground-based site operations are fully described in Rajot et al. (2008).

The aircraft observations were conducted from Niamey, Niger. The wintertime airborne campaign, called AMMA SOP0-DABEX, was conducted onboard the Facility for Airborne Atmospheric Measurements (FAAM) BAe-146 aircraft (Haywood et al., 2008). The aircraft performed thirteen research flights in the area between 8°N and 18°N, including three dedicated excursion to the north of Niger, toward the Air Mountains and the Ténéré desert to probe specifically mineral dust whilst transport. In summertime, the observations were conducted onboard the Service des Avions Français Instruments pour la Recherche en Environnement (SAFIRE) ATR-42, which operated twenty research flights in the area between 6° and 20°N, therefore crossing the ITCZ to the north to contrast the Sahara and the Sahelian dust properties in their respective boundary layers (Reeves et al., 2010; Formenti et al., 2011b).
The Geostationary Earth Radiation Budget Intercomparison of Longwave and Shortwave radiation (GERBILS) found its scientific motivation in the discrepancy that had been found between the satellite observations and the numerical weather model predictions of the top of the atmosphere outgoing longwave radiation over desert areas in western Africa (Haywood et al., 2005). This discrepancy could be reconciled by including mineral dust in the model. The GERBILS experiment, whose full rationale and operating details are described in Haywood et al. (2011), was therefore designed to test this hypothesis, and to determine the properties of the mineral dust to be included in the model. The experiment consisted in ten research flights onboard the FAAM BAe-146 performed between Nouakchott (18.10°N, 15.94°W, Mauritania) and Niamey (13.48°N, 2.18°E, Niger) on an almost standard route along the 18°N meridian. Flights were performed between 18 and 29 June 2007, in a period characterised by a persistent and widespread dust load.

Finally, the Dust Outflow and Deposition to the Ocean (DODO) airborne experiment took place at two different periods in 2006: in wintertime (7–16 February) and in summertime (22–28 August). The aim of DODO was to characterised dust as a possible nutrient for the marine ecosystems. For this reason, the FAAM BAe-146 aircraft was based in Dakar, in Senegal, and most of the twelve research flights (six in the winter period and six in the summer period, respectively) were conducted over the Atlantic Ocean in the outflow from the African continent. However, two flights, one per season, were performed inland above Mauritania in order to sample dust at or very close after emission (McConnell et al., 2008).

2.2. Sample collection and handling

2.2.1. Ground-based

Aerosol sampling at the ground-based AMMA super site of Banizoumbou was performed using two identical purpose-built wind-oriented inlets designed for the AMMA field campaigns (Rajot et al., 2008). The cut-off diameter 50% efficiency of these inlets has been calculated to be approximately 40 µm by using the standard formulae of particle losses in inlets and tubings reported in Baron and Willeke (2001) and Hinds (1999). Each inlet leads to a sampling chamber containing seven different sampling lines, each of them dedicated to a different instrument. Multiple sampling lines are very convenient as they allow collecting various filter samples in parallel, differentiated and optimized in terms of sampling medium, exposure interval and flow rate as a function of the analytical technique to be applied. This also results in minimal manipulation of the filters after sampling. Samples were collected on acid washed 42-mm polycarbonate Nuclepore filters (0.4 µm pore size) mounted on plastic rings. Samples were stored in Petri dishes after sampling. Sampling time was of the order of hours, and was varied depending on the aerosol mass concentration measured on-line by a
Tapered Element Oscillating Microbalance (TEOM, Rupprecht and Patashnick, Albany, USA) (Rajot et al., 2008).

### 2.2.2. Airborne

During the AMMA SOP0-DABEX, DODO and GERBILS campaigns, samples were collected onboard the FAAM BAe-146 aircraft. The aircraft is capable of carrying 2 crew, 18 scientists and a total scientific payload of up to 4000 kg for a distance of 3700 km with a ceiling of 35,000 feet and has a typical science speed of 110 m s\(^{-1}\). For the typical scientific payload, flight patterns, ambient meteorological conditions and the proximity of diversion airports encountered during the campaigns, the average endurance of the BAe146 aircraft was approximately 5 hours. Sampling was been performed using a set of two parallel thin-walled inlet nozzles with a curved leading edge; the design was based on criteria for aircraft engine intakes at low Mach numbers (Andreae et al., 1988). The sampling system was operated at flow rates that averaged at 120 L min\(^{-1}\) (at ambient pressure and temperature; the flow was adjusted to maintain slightly subisokinetic sampling conditions. The passing efficiency of the inlets has not been formally quantified. During AMMA SOP-0/DABEX, Chou et al. (2008) showed that the number size distributions of the aerosols collected on the filters (counted by electron microscopy) extended up to 10 µm diameters and were of comparable magnitude to those measured by wing-mounted optical counters.

Aerosol particles were sampled by filtration onto two stacked-filter units (SFUs) mounted in parallel. Each SFU can hold a maximum of three filters on sequential 47- or 90-mm diameter polyethylene supports. Only one stage was used during the campaigns, i.e., samples represented the aerosol bulk composition. Samples were collected only during horizontal flight legs lasting not less than 20 min in order to guarantee sufficient loading of the filter samples. As a consequence, at the aircraft cruise speed, each sample had at best a spatial resolution of 180 km. Each SFU consisted of a Nuclepore filter (Whatman) of nominal pore size 0.4 µm.

During the AMMA SOP1-2 campaigns, samples have been collected onboard the SAFIRE ATR-42. With the necessary payload for the campaign, the aircraft had a maximum endurance of 4 h. Minimum flight altitude is 300 m over land. Ceiling is approximately 7 km. The aircraft was equipped with basic sensors for measuring the radiative, dynamic and thermodynamic properties of the atmosphere (Saïd et al., 2010).

Aerosol sampling was performed using the AVIRAD aerosol sampling system newly developed for AMMA (Formenti et al., 2011b). It consists of an iso-axial and isokinetic inlet whose 50% passing efficiency has been estimated at 9 µm in diameter. At the cruise speed of the ATR-42 (93 m s\(^{-1}\)) the delivered volumetric flow rate is 350 L min\(^{-1}\), which was
distributed to various instruments, including two SFU for collecting bulk aerosol samples. Samples were obtained by parallel filtration onto 42-mm diameter polycarbonate membranes (nominal pore size 0.4 µm Nucleopore, Whatman) on polyethylene supports. Samples were collected only during horizontal flight legs lasting not less than 20 minutes. Immediately after each flight, the loaded filters were stored in Petri dishes.

Ground-based operation with multiple sampling lines allowed for parallel sampling where the sampling duration is adapted to the characteristics (detection limit, saturation level) of the technique to be employed for analysis. On aircraft, the two parallel samples available had to be manipulated further to allow accommodating the number of analysis one wanted to perform, also taking into account the fact that some analyses are destructive of the sample. Therefore, the handling protocol for aircraft samples consisted in analysing entire filters or portions of those depending on their load on mineral dust. A dedicated tool in polyethylene has been developed for cutting the 47- or 90-mm filters into halves, quarters or eighth without touching and therefore contaminating them.

2.3. Sample analysis

2.3.1. Elemental composition

The total elemental concentrations for elements from Na to Pb by wavelength dispersive X-ray fluorescence (WD-XRF) for samples collected on the ground and on the ATR-42 and Particle-Induced X-ray Emission (PIXE) for the aircraft samples collected onboard the BAe-146. The reason for using different techniques for ground-based and aircraft samples lays on the fact that the WD-XRF sample holder set up could not, at the time, host filters which are not supported on polyethylene frames. The details of the analytical protocols are provided in Formenti et al. (2008; 2011b), McConnell et al. (2008), Rajot et al. (2008) and Klaver et al. (2011). Elemental concentrations are estimated with a 5% error. An inter-comparison exercise on samples generated in the laboratory from geo-standards of known and certified composition has shown that the two techniques yield comparable results within the error bars (Formenti et al., 2010).

2.3.2. Mineralogical composition

The identification of major minerals composing mineral dust: quartz (SiO$_2$), calcite (CaCO$_3$), dolomite (CaMg(CO$_3$)$_2$), gypsum (CaSO$_4$•2H$_2$O), clays (kaolinite, illite, smectite, chlorite), feldspars (orthoclase KAlSi$_3$O$_8$ and albite NaAlSi$_3$O$_8$) was performed by X-ray diffraction (XRD) analysis at the Institut de Recherche pour le Développement (IRD) in Bondy, France. The diffractometer, a Siemens D500 with Ni-filtered Cu-K$_\alpha$ radiation, was operated at 40 kV and 30mA. Samples were scanned from 2 to 70° (2θ) with counting for 10 s every 0.02° (2θ). The analytical procedure and semi-quantitative treatment are fully described by Caquineau et
al. (1997), who adapted the sample preparation to low-mass mineral aerosol (load deposited on filter > 800 µg). The calibration protocol that was applied to the XRD spectrometer in order to quantify the mineralogical composition is described in Klaver et al. (2011). An improvement of the calibration curves was performed after publication of the Klaver et al. (2011) paper by achieving the calibration for the feldspars (orthoclase and albite). The calibration factors $K_i$ representing the ratio between the peak surface area $S_i$ in the diffraction spectra and the mass $m_i$ of the i-mineral are reported in Table 1. The error on the mineral masses is estimated as the sum of the relative error on the peak area in the diffractograms and on the calibration factor $K_i$ as $\frac{\Delta m_i}{m_i} = \frac{\Delta K_i}{K_i} + \frac{\Delta S_i}{S_i}$. The term $\Delta K_i/K_i$ is estimated from the error of the slope of the calibration curve (Table 1) and it varies between 2% and 10%. The term $\Delta S_i/S_i$ depends on the abundance of the mineral in the samples and has been estimated from the obtained counting statistics.

2.3.3. Iron oxide content and speciation
The iron oxide content, defined as the fraction of iron which is not in the crystal lattice of silicates (Karickhoff and Bailey, 1973), was determined with the adapted Citrate-Bicarbonate-Dithionite (CBD) extraction method developed by Lafon et al. (2004). This method is an adaptation for aerosol filters (with typical material mass less than 500 µg) of the classical method of (Mehra and Jackson, 1960) for soil analysis. The method uses a reagent to dissolve iron oxides selectively via reduction. Details of this analysis are given in Formenti et al. (2008; 2011b) and Klaver et al. (2011). The percent error on the iron oxide content is estimated to be 10% as its assessment is based on two XRF analysis (prior and after dissolution of the iron oxides from the sample).

The speciation of iron oxides was performed by X-ray absorption (XAS) at the Fe K-range: XANES (X-ray Absorption Near Edge Structure) and EXAFS (Extended X-ray Absorption Fine Structure) at the SAMBA (Spectroscopies Applied to Materials based on Absorption, http://www.synchrotron-soleil.fr/portal/page/portal/Recherche/LignesLumiere/SAMBA) line SOLEIL synchrotron facility in Saclay, France. XAS spectroscopy is based on the analysis of the position and shape of the K-pre-edge and edge peaks, depending on the oxidation state of iron but also to the atomic position of the neighbouring atoms, mostly O+ and OH-. The quantification of the Fe-status was based on the analysis of five standards of Fe(III)-bearing minerals which can be found in mineral dust. Full details of the experimental protocols and the data analysis are provided in Formenti et al., Dominance of goethite over hematite in iron oxides of mineral dust from Western Africa: quantitative partitioning by X-ray Absorption Spectroscopy, submitted to J. Geophys. Res., 2014 (hereafter referred as Formenti et al., 2014).
2.4. Data interpretation and presentation

2.4.1. Source identification

The identification of source regions has been performed using a variety of ancillary products:

(1) The record of measurements of the horizontal movement of sand grains by a saltiphone (Eijkelkamp, Giesbeek, The Netherlands) located at the proximity of the sampling inlets at the super-site of Banizoumbou providing with the estimate of the saltation horizontal flux related to established local erosion (Sow et al., 2009).

(2) Various satellite products, in particular, the daily images of the UV aerosol index by the Ozone Monitoring Instrument (OMI) on Aura; the instantaneous dust product maps from the SEVIRI satellite; and the Infrared Difference Dust Index (IDDI) obtained from the infrared channel of Meteosat (10.5-12.5 μm) (Legrand et al., 1994).

(3) Back-trajectories and air mass dispersion calculations using the Met Office Numerical Atmospheric-dispersion Modeling Environment (NAME, Cullen, 1993) and the NOAA HYbrid Single-Particle Lagrangian Integrated Trajectory Model (HYSPLIT) as described in Chou et al. (2008), Rajot et al. (2008) and Klaver et al. (2011).

2.4.2. Data presentation

2.4.2.1. Calculation of the total mass

In the following, the composition of mineral dust will be presented as percent of the total dust mass (TDM). According to Pye (1987), the TDM can be decomposed as

\[
TDM = m_{\text{clays}} + m_{\text{quartz}} + m_{\text{Ca-rich}} + m_{\text{feldspars}} + m_{\text{iron oxides}} + m_{\text{tit oxides}} + m_{\text{organic}} \quad (1)
\]

where

\( m_{\text{clays}} \) = mass of the major clay-species (kaolinite, illite, smectite, chlorite)

\( m_{\text{quartz}} \) = mass due to quartz (SiO\(_2\))

\( m_{\text{Ca-rich}} = m_{\text{calc}} + m_{\text{dolom}} + m_{\text{gypsum}} \) = mass due to calcium carbonates and sulphates in the form of calcite (CaCO\(_3\)), dolomite (CaMg(CO\(_3\))\(_2\)) and gypsum (CaSO\(_4\))

\( m_{\text{feldspars}} \) = mass due to feldspars, such as albite and ortoclace (NaAlSi\(_3\)O\(_8\) and KAlSi\(_3\)O\(_8\), respectively)

\( m_{\text{iron oxides}} \) = mass due to hematite (α-Fe\(_2\)O\(_3\)) and goethite (α-FeOOH)

\( m_{\text{tit oxides}} \) = mass due to titanium oxides (TiO\(_2\))
$m_{\text{organic}}$ = mass due to organic matter which is present in the soils as biological debris and complex organic molecules (humus).

In practical terms, the TDM can be approximated by the TEDM (total estimated dust mass) as the sum of the oxide-equivalent of the elemental concentrations of the major elements in mineral dust (Al, Si, Fe, Ca, K, Ti accounted as Al$_2$O$_3$, SiO$_2$, Fe$_2$O$_3$, CaCO$_3$, K$_2$O and Ti$_2$O) whose masses are measured by XRF analysis. The TEDM estimated this way can be compared to the direct measurement of the total dry gravimetric mass (TDGM), which is available for some of the datasets (ground-based AMMA SOP0 and SOP1-2, and GERBILS), either by weighing or by on-line Tapering Element Oscillating Microbalance (TEOM, model 1400a, Rupprecht and Patashnick Albany, New York USA) (see Rajot et al. (2008) and Klaver et al. (2011)).

The slope of the least-squared linear regression between the TEDM and TDGM, calculated on the 81 samples on which the extraction of iron oxides was performed (see section 3.3), is 0.98 and the coefficient of determination ($R^2$) is 0.88. This is very satisfactory, taking into account that, because the mass is dominated by coarse particles, which have large mass but little number, differences might arise by losses of some coarse particles between the time at which the filter was weighed and the time at which it was analysed by XRF or by small differences between the filter sampling duration and the TEOM integration times. As a matter of fact, the most of the scatter in the data is observed for samples which corresponded to local erosion or transported emission by convective systems, having the largest fraction of coarse particles and the shortest integration times. This becomes apparent when plotting separately the ground-based AMMA SOP0 and SOP1-2 data (Figure 1).

Therefore, we consider that the TDM is equal to the TDGM for samples for which the TDGM is measured. When this is not available, the TDM is approximated via the TEDM calculated as sum of oxides. The percent uncertainty is calculated as the squared sum of the errors on the calculation of the TEDM from the measured Al, Si, Fe, Ca, K, and Ti elemental masses. By doing so, the percent uncertainty on the TDM is 12%.

### 2.4.2.2. Calculation of the mineralogical composition

The mass apportionment presented in Equation 1 has been performed as follows:

(1) the masses of quartz, calcite, dolomite, gypsum and feldspar ($m_{\text{quartz}}, m_{\text{Ca-rich}}, m_{\text{feldspars}}$) are quantified by XRD using the mass/intensity calibration curves as described in section 2.3.
The mass of iron oxides ($m_{\text{iron oxides}}$) is measured directly by CBD analysis. The apportionment of $m_{\text{iron oxides}}$ into its main components (hematite and goethite) is performed by XANES spectroscopy.

The mass of the clay species, kaolinite, illite, smectite, chlorite ($m_{\text{clays}}$) is estimated by the difference between the TDM and sum of the masses of $m_{\text{quartz}}$, $m_{\text{Ca-rich}}$, $m_{\text{feldspars}}$ and $m_{\text{iron oxides}}$. The apportionment of $m_{\text{clays}}$ into its constituting species is difficult to achieve without a direct calibration. This is highly uncertain as there is basically no mineralogical reference having the same crystallographic and chemical status than in the actual aerosol samples, because of the exposure to the environment, weathering in particular, that they experience before and after become airborne. Furthermore, smectite clays are actually a family of species with different chemical characteristics (see the general chemical formulae in Appendix 1 for montmorillonite and nontronite). When only illite and kaolinite are present in the samples, and no traces of chlorite or smectite are shown by XRD analysis, their mass can be apportioned readily using the illite-to-kaolinite ratios ($I/K$) established by Caquineau et al. (2002) as a function of dust source region as

$$m_{\text{kaolinite}} = \frac{m_{\text{clays}}}{1 + \frac{I}{K}},$$

$$m_{\text{illite}} = m_{\text{clays}} - m_{\text{kaolinite}}$$

When smectites or chlorite is also detected, Equation (2) does not hold true. We will evaluate in the following sections the uncertainties induced by this approximation.

The mass of titanium oxides $m_{\text{tit oxides}}$ is estimated by calculating the TiO$_2$ contribution as the measured elemental Ti multiplied by 1.67, the ratio between the atomic weight of TiO$_2$ and that of elemental Ti;

The mass of organic compounds ($m_{\text{organic}}$) in mineral dust was generally neglected on the basis of the work by Lepple and Brine (1976), it should not exceed 3% of the TDM.

We also did not take into account the mass of diatomite fragments which have been observed in samples originated from Bodélé (Chou et al., 2008). Diatomite fragments are silicate skeletons of algae and they do not have a crystalline structure which can be quantified by XRD analysis. Their presence can be put into evidence by elemental analysis as excessive values of the Si-to-Al ratio, up to 4 (Formenti et al., 2011b). The
3. Results

The whole dataset (ground-based and airborne AMMA SOP0 and AMMA SOP1-2, GERBILS and DODO campaigns) consists in 704 samples for which we measured the total elemental composition. Out of those, 54 samples have been selected to yield data on the major mineral composition, 86 have been analyzed to yield the content of iron oxide, 12 to determine the iron oxide speciation. 31 samples combined sufficient information to determine the full mineralogical composition. The sample selection has been based on the filter loading and on the possibility of identifying their provenance and attributing them a source region.

3.1. Identification of source regions

The identification of the origin of the mineral dust collected on the samples implies being able to make the link between the knowledge on the position and the mineralogy of the source regions to that of the aerosol samples. To do so, one should know (1) the transfer function, at emission, linking the soil mineralogical/chemical composition to that of the aerosols, and (2) the transfer function describing how its composition is modified during transport.

The first transfer function describes the mineralogical fractionation that occurs between the soil and the aerosol as a consequence of the size segregation happening when the soil grains blast on the soil after saltation (Rahn, 1976). The fractionation can alter the elemental ratio between Si and Al, as a result of the depletion in quartz and feldspars with respect to clays at emission. Quartz and feldspars have higher Si content than clays and are more abundant in the sand than in the silt/clay fractions of the soil (Mason, 1966; Rahn, 1976; Chatenet et al., 1996). These last are enhanced during emission by the saltation/sandblasting (Alfaro et al., 1998). Whereas there is still no systematic assessment of the transfer function between the soil and the aerosol composition, Claquin et al. (1999) have shown that some mass ratios between minerals (for examples, that of illite to kaolinite and that of calcite to quartz) are conserved from the soil to the aerosol if the comparison is restricted to roughly the same size fractions (clays or silt). Lafon et al. (2006) also suggested that the free-iron fraction to the total is conserved. Jeong (2008) has shown that the calcite and the K-feldspar fractions are very similar in Asian dust and the corresponding silty soil
fraction. The percent of phyllosilicate is higher in the dust aerosols than in the soil, whereas the quartz and plagioclase fractions are lower.

The second transfer function describes how the dust composition is modified during transport. Our dataset consists in data which have been collected at most 2 days after emission. We expect that only losses of particles of diameter larger than 10 µm would occur in this short time scale (Pye, 1987). We therefore assume that the chemical/mineralogical composition of our dust aerosol samples is linkable to that of the parent source regions.

To identify the pathway of air masses during transport, and trace them back to an active dust source, we used back-trajectory and dispersion modelling and various satellite products (e.g., Chou et al., 2008; Klaver et al., 2011). This approach does not apply to samples collected in the boundary layer during the Monsoon season south of the Inter-Tropical Convergence Zone (ITCZ). In this case, the dominant flow is from the south-west: dust is present in the boundary layer only as a result of erosion by organized convection or under particularly stable conditions, when high temperatures establish and the thermal turbulence in the boundary layer raises its top height at or above the shear level, bringing dust from the free troposphere in the surface level (Flamant et al., 2009; Formenti et al., 2011b).

Even when applicable, this approach has various caveats: first, if their horizontal velocity is high and exceed the local threshold velocity, air masses within the boundary layer might get loaded with dust along their way before reaching the sampling point. This happens for example when emission is induced by large scale frontal systems as the one of 3–7 March 2006 (Tulet et al., 2008). Secondly, ground-based sampling lasts several hours and the direction of air masses might change in between. Finally, aircraft samples integrate over large areas to which different air masses from different source regions might contribute. This problem might be overcome, or at least minimized, by increasing as much as possible the sampling frequency to match the variability of transport. We did that by following the variability of concentrations, in time or space, with ancillary measurements at higher temporal resolution (mass concentration by the TEOM balance, or the scattering coefficient by a TSI nephelometer).

In spite of this, we cannot exclude that our samples might represent larger emission areas than single "hot spots" (see for example, Figure 1 in Klaver et al. (2011)). We do not necessary regard this as a problem, as the atmospheric dust load, both at short and large-scale transport, is often due to a variety of sources active at the same time (Glaccum and Prospero, 1980; Reid et al., 2003; Formenti et al., 2003; Schepanski et al., 2007; Laurent et al., 2008; McConnell et al., 2008; Marticorena et al., 2010). Also, this supra-imposition, both in the vertical and in the horizontal, is what is actually seen by spaceborne sensors.
results of this paper might then be used to build a dust composition model on a scale
relevant to satellites.

When doing so, we found that samples presented in this paper are representative of major
African sources which, according to Formenti et al. (2011) and Scheuvens et al. (2013), are
indicated as PSA (Potential Source Area) in Figure 2.

- The PSA2, including arid areas in Western Sahara, Mauritania and Morocco, sampled
during the DODO and the GERBILS campaigns. With respect to the extent
of this source, the northern part is under-represented as it was sampled on one
occasion only;

- The PSA3 represents emitting area in northern Mali and southern Algeria. This is
largely the dominant source in terms of frequency of sampling. The air masses
travelling from PSA3 often show a recirculating pathway over North Niger. By doing
so, at times they might pick up dust from the desert areas close to the Aïr massif,
which is not depicted in Figure 2, before reaching the sampling site. These
occurrences, when identified, are labelled as PSA3-NN;

- The PSA4 was sampled on a few occasions during wintertime. The PSA4 was
sampled on a few occasions during wintertime. As for PSA3, samples originating from
PSA4 for which the air masses had got in contact with the surface in northern Niger
when satellite images indicated emissions, have been labelled as PSA4-NN;

- A few samples also represent dust originated from the Bodélé depression (PSA5). In
summertime, a mixture of Sudan and Chad sources including Bodélé might have
been sampled (Flamant et al., 2009). Finally, a number of episodes of erosive
emissions by convection in the Sahel were also sampled and characterised
(referenced as Sahel). These occurred both over Mali and over Niger.

The PSA1 source is not represented in this paper as it was not sampled during the field
campaigns.

To provide with a first link between geographical origin and chemical composition, we traced
the scatterplot of elemental Si/Al versus Fe/Ca which we color-coded according to the source
region according to satellite/back-trajectories/dispersion modelling information. These ratios
have been shown previously to be rather robust indicators of the origin of mineral dust at the
large scale (Formenti et al., 2011a and references therein). The result is shown in Figure 3.

There are some clear distinctions which that can be made. Data points having Si/Al > 3 are
unambiguously identified as samples from the Bodélé depression, i.e. PSA 5. Their Fe/Ca
ratio is about 1. Data point having Fe/Ca > 3 are related to emission from erosive events in the Sahel. There are various sub-ranges in this group: data having Fe/Ca in the range 3–5 correspond to transported dust aerosol from the Sahel, sampled 1–2 days after emission, corresponding to mixed sources, as PSA3 + Sahel. Data having Fe/Ca > 5 correspond to samples collected immediately or at the very early stage after emission by convective events, mostly locally at the ground-based Banizoumbou site (when the Fe/Ca ratio exceeds 10), but occasionally on uplifted dust onboard aircraft.

Data with Si/Al in the range 2.5–2.9 and Fe/Ca in the range 1–3 have their origins mostly at latitudes in the range 16–27°N, i.e. PSA 3. Whereas the variability of the Si/Al ratio is low, that of Fe and Ca can be used to make some distinctions amongst sources. Fe/Ca higher than 1 is measured for data points from northern Mali, Mauritania, Algeria, northern Niger and Libya whereas Fe/Ca < 1 is for dust from Western Sahara (PSA 2). For this source, there are two clear outliers corresponding to samples of dust originating from Morocco and showing the lowest Fe/Ca ratio (~ 0.4).

3.2. Mineralogical composition

The investigation of the mineral composition is based on 51 samples, 15 of which collected onboard aircraft. This dataset represents a diversity of sources and emission conditions. The most evident peaks, for all samples, are those of illite, kaolinite and quartz. Less evident, and not ubiquitous, are the peaks corresponding to feldspars (albite and orthoclase), and calcium-rich minerals (calcite, dolomite, gypsum). Feldspars are found in the samples from local erosion and South Algeria/North Niger/Mauritania and Western Sahara. This last set might also contain calcium-rich minerals. Smectite clays are detected as a large and rather weak peak at angles comprised between 4 and 7°. This is indicative of the fact that smectites are a family of weathered clays of different chemical composition (e.g., montmorillonite, nontronite), whose crystalline structure can be heavily modified in the environment. Smectite clays were only evident in samples originated from South Algeria/North Niger/Mauritania and Western Sahara.

We can obtain the mass fractions of minerals which can be detected by X-ray diffraction by using the calibration factors as described in section 2.3.2 (Figure 4). Regardless of the origin, clays and quartz account in average for 87% (± 6) and 10% (± 6) of the diffracting mass. There is only one clear outlier in an episode of intense local erosion sampled at the Banizoumbou super-site (sample SOP1-8), for which the percent clay fraction is reduced (56%), and the quartz and feldspars fraction enhanced (41% and 3.3%, respectively). This is consistent with the fact that at erosion the size distribution shows an
enrichment of coarse particles, whose composition should be closer to that of the soil silt fraction (Gillette and Walker, 1977; Chatenet et al., 1996).

The analysis of the XRD spectra also show that kaolinite and illite are very often the only clay species detected. A broad peak corresponding to smectites was detected on samples from Algeria and at times Bodélé, consistently with finding of Moreno et al. (2006). No chlorite has been detected in our samples. As explained in section 2.4.2.2, the relative proportions (by mass) of illite and kaolinite have been estimated using the origin-dependent ratios established by Caquineau et al. (2002). For the sources considered in this dataset, the illite-to-kaolinite ratios could vary between 0.1 and 2.3. The central and western Saharan sources (PSA2 and PSA3) display the highest values (1.6 to 2.3) for PSA 5, Northern Niger and Sahel the lowest values (0.1 to 0.7).

### 3.3. Iron oxides content and speciation

The quantification of the content of iron in the oxide form (Fe$_{ox}$), that is free iron which is not in the crystal lattice of alumino-silicates, has been performed on 81 samples of different origins and load (33 out of those have been collected onboard aircraft). In average, Fe$_{ox}$ account for 58 ± 7% of the mass of elemental Fe. The range of variability is illustrated in Figure 5.

The lowest values (down to 44%) are found for samples of dust originating from the PSA4 (Chad/Egypt) area and for a set of samples collected above the PSA2 areas. The highest values (up to 71%) are obtained for samples of dust emitted by local erosion in the Sahel, including one episode in which uplifted dust has collected onboard aircraft after emission. Other subsets of samples, such as those of dust originated from the South Algeria/North of Niger areas, display equivalently high values when during long-range transport, dust from Algeria mixes with locally-produced dust in Niger. It should be noted also that, because of the internal variability of each group, there is no statistically-significant difference between the iron oxide fractions of the total iron for long-range transported dust originated at latitudes north of 16°N. A broad distinction between the Fe$_{ox}$-to-Fe ratio of Sahelian samples with respect to those of Saharan samples can be made by sorting the data points by their corresponding Fe/Ca ratio. Mineral dust emitted from the Sahel (Fe/Ca > 3) are characterized by a mean Fe$_{ox}$-to-Fe ratio value of 65% (± 5%, standard deviation of the mean), whereas episodes of transport from the Sahara (Fe/Ca < 3) are characterized by a mean ratio of 56% (± 6%). No further distinction based on origin is possible in this second group of samples. This is illustrated in Figure 6.

X-ray absorption spectroscopy on twelve samples of differing origins indicated that goethite is overall the dominant species of iron oxides. Goethite accounts for 48 to 73% of the iron
oxide mass, whereas hematite accounts for 27 to 52%. These values do not show a clear
trend with respect to the sample origin. The relative proportions of iron in the form of
hematite and goethite with respect to total iron are shown in Table 2.

3.4. Titanium oxides content

Titanium oxides deserve attention as they are involved in photo-catalytical heterogeneous
reactions with various atmospheric constituents, including volatile and semi-volatile organic
compounds which are found in transport regions of mineral dust due to biomass burning
emissions (Gustafsson et al., 2006).

We therefore examined our dataset in order to provide with estimates of the content of
titanium oxides in mineral dust from various sources in western Africa. The titanium oxide
content has been estimated from the measured element Ti concentration (obtained by XRF)
which is converted to TiO$_2$ using the appropriate molar mass conversion factor (1.67). This
should be considered as an upper limit estimate as some of the titanium in mineral dust is
present as substitution element in aluminosilicates (Dolcater et al., 1970).

Figure 7 presents the correlation between the TiO$_2$ and the TDM. Two populations can be
identified: one corresponding to samples collected in correspondence of erosion events and
one corresponding to advection events. For both populations, the correlation is excellent ($R^2$
equal to 0.97 and 0.92, respectively) and the scatter of points is minimal. The TiO$_2$ content is
higher in erosion events from the Sahelian region (average 1.2% ($\pm 0.1$) of the mass), than
for advection events of Saharan dust, for which TiO$_2$ represents in average 0.71% ($\pm 0.01$) of
the mass.

Looking at the elemental ratios between Ti, Fe, and Fe$_{ox}$ provides with further insights
(Figure 8). The ratio of Ti-to-Fe is insensitive to changes in the Fe$_{ox}$ proportions to Fe up to
60%, and then starts increasing linearly with increasing Fe$_{ox}$-to-Fe ratio. This suggests that,
in Sahelian dust, Ti is associated with iron oxides but not in Saharan dust.

3.5. Mass apportionment of the mineralogical composition

By combining the information presented in section 3.1 to 3.4, we can estimate the
mineralogical composition of samples of mineral dust of local or distant origin.

The mass apportionment is shown in Table 3. Note that we provide here with the
composition of the bulk mineral dust, without explicitly taking into account the dependence on
size.

There are clear similarities in the composition of the dust from those different origins, first of
all in the fact that the clay fraction represents the largest fraction of the dust mass. Also note
the absence of Ca-rich minerals (calcite, dolomite, gypsum) in dust originating from the Sahel compared to that originated elsewhere.

Despite the overall similarities, there is indeed a large internal variability within each source group, expressed by the standard deviation around the mean, and which might be related to either the time or height after transport.

An example can be given by looking at the composition of the four samples in Bodélé/Sudan group: one sample was collected in the dry season, whereas the remaining three are sequenced samples collected in the wet season, at a time characterised by the absence of local emission, but by a progressively increasing entrainment of dust transported from the free troposphere to the surface layer (J. L. Rajot, pers. comm.). These wet-season samples have a lower clay fraction and a higher quartz and feldspar content (in the range 7–18% and 0.4–0.8, respectively). This variability corresponds to differences in the volume size distribution, normalised to the total volume for sake of comparison (Figure 9). As a matter of fact, the volume size distribution of the dry-season sample presents a large coarse mode peaking around 4–5 µm, whereas the wet-season samples have a bimodal size distribution with an additional mode centred at 8 µm. Within this group, the sample having the largest quartz fraction is also the one displaying the larger fraction of particles above 10 µm.

A second example of the links between time after emission, size distribution and composition, is given in Figure 10 showing the normalised particle size distributions (by volume) for one case of Sahelian dust by local erosion (sample SOP1-8), two samples representing the background dust composition in the absence of local erosion (SOP1-15 and SOP1-60) and one case representing the composition of Sahelian dust advected at the sampling site after having been emitted by local erosion in the proximity (SOP1-17). There are clear differences between these cases: sample SOP1-8 has a mode at 8 µm and the largest coarse particle fraction. Samples SOP1-15 and SOP1-60 have a bimodal size distribution with modal diameters at 3–4 and 8 µm. Finally, only the mode 3–4 µm remains for sample SOP1-17. As for the case presented in Figure 9, it is evident from Figure 10 that the coarse mode at 8 µm bears the most of the quartz and feldspar mass fractions, and that the dynamic evolution of this mode controls the content of quartz and feldspar in dust. The comparison of Figures 9 and 10 also suggests that, once the mode at 8 µm is deposited, differences in the mineralogical composition of mineral dust will be more independent of the size dynamics and reflect better differences in the mineralogy of the parent soil.
4. Discussion

4.1. Comparison with other datasets

All comprehensive papers relating on the composition of mineral dust state that the comparison of compositional data from different datasets depends critically on the experimental conditions, in particular the duration and location of sampling (integration time, near-source location) (Formenti et al., 2011a; Scheuvens et al., 2013). These considerations are particular relevant to aircraft sampling where the difficulty in controlling the size cut-off of samplers generally disfavour the coarse particle fraction in an often non-quantified way (Formenti et al., 2011a). Furthermore, as the aircraft is moving during sampling, a 20-minute aircraft sample will represent a distance of the order of 120 km, which might integrate the contribution of various sources (Klaver et al., 2011).

As a consequence, the integration and the comparison of different datasets of the composition of mineral dust is not a straightforward exercise.

Overall, our results indicate the dominance of clays in the dust composition when only the aerosol fraction is measured. In this respect, they are in accordance with previously published papers for dust collected in Africa (Kandler et al., 2009) or after long range transport (Delany et al. 1967; Kiefert et al., 1996; Glaccum and Prospero, 1980; Caquineau et al., 2002). These measurements, however, tend to indicate that the clays fraction accounts for ~60-70% of the dust weight, whereas our measurements suggest higher values (80-90%). It should be reminded that the determination of the mineralogical composition will depend on the crystallography of the standard minerals used for calibration, and on the normalisation to the total dust mass. In our work the percentage of clays in dust depends directly on the normalisation factor, on which we have an uncertainty of the order of 17%, as we do estimate clays from the difference between the total dust mass and the calibrated mass of non-clay compounds. As an example, Glaccum and Prospero (1980), using standard clays minerals for calibration, estimate that the crystalline mass represented 70 to 110% of the weighed mass. These values reflect, at least partially, the uncertainty on the clay fraction.

Nonetheless, in some cases differences might be due to sampled aerosol size distribution. As a matter of fact, the lowest detected percentage of clays corresponds to sample SOP1-8, collected during an intense episode of local erosion. In this case, clays account for around 50% of the measured mass, in accordance with measurements by Caquineau et al. (1997; 2002) in similar conditions.

This is also clear for example by looking at the data published by Kandler et al. (2009), who documented the dust composition in dust storm (total suspended particulate concentrations higher than 300,000 μg m⁻³) and low dust periods (total suspended particulate
concentrations around 100 µg m$^{-3}$) in Morocco. Quartz accounts for as high as ~65% in dust storm conditions, suggesting that soil grains, and not only aerosols, were sampled. Conversely, low-dust conditions are characterised by only about 20% of quartz. These authors showed that, for locally emitted dust, illite is the dominant clay form over kaolinite (up to 30 and 5% by volume, respectively). These authors also detected clay in the form of chlorite (less than 10% by weight) and calcite (up to 15%).

Few authors have published data on the iron oxide content. Lafon et al. (2004) evaluated the free-to-total iron content in samples collected at the Banizoumbou ground-based site and found mean values of 0.44 (± 0.11) and 0.65 (± 0.04) for dust transported in the Hamattan flow (from the Sahara) and dust emitted locally by convective erosion. Lazaro et al. (2008) found sensibly lower values than ours ranging from 0.26 to 0.63 but only exceeding 0.56 in three occasions when back-trajectories indicated transport from the Sahel or a mixture of dust originating from the Sahara and the Sahel. Our values are on the higher side with respect to those of these authors. Nonetheless, the back-trajectories presented in the paper seem indicating that the Saharan dust originated from higher latitudes that the one we sampled – and mostly from Tunisia and Algeria. It will be interesting in the future to sample those source areas to confirm these rather low values.

Finally, our results on the iron oxide speciation are consistent with findings of different authors using XANES speciation (Wilke et al. 2001; Ohta et al., 2006; Prietzel et al. 2007; Majestic et al., 2007) and visible spectroscopy (Arimoto et al., 2002) indicating that Fe(III) is the dominant red-ox form of iron oxides and that goethite is the predominant form of iron oxide over hematite.

### 4.2. Atmospheric implications

The results showed in this paper suggest that, when restricted to the mass fraction below 20 µm, the dust composition in western Africa may apparently be simplified to few minerals, clays, quartz, calcite, iron and titanium oxides. Feldspars and other calcium-containing minerals such as dolomite and gypsum are very minor components close to detection limits. The geographical distribution of the dust composition is relatively unimportant in terms of mass fractions, in particular in comparison to the importance of the size distribution whose variability with time seems to be able to explain some differences observed between samples from the same source region at different times after emission. Nonetheless, it is important to evaluate whether differences in the composition at the regional scale, even when minor, might give raise to differences in the impacts of mineral dust.

To do so, we explore the influence of dust composition on the spectral complex refractive index $\bar{n}$ and the fraction iron solubility.
4.2.1. Implications for optical properties

As a mean of evaluating the effect of the variability of the composition on the dust optical properties, the volume-average refractive index has been calculated as

\[ \tilde{n} = \sum_j f_j \times \tilde{n}_j \quad (4) \]

where \( f_j \) is the volume fraction of each individual mineral in the dust sample (as reported in Table 3) and \( \tilde{n}_j \) is its complex refractive index. Equation 4 assumes that minerals in dust are in internal mixing. In reality, minerals are present in dust as external mixtures, with the exception of iron oxides which may be found as inclusions in clays (Sokolik and Toon, 1999).

However, the internal mixing hypothesis is frequently used in experiments and climate models (e.g., Balkanski et al., 2007; Kandler et al., 2009; McConnell et al., 2010; Hansell et al., 2011).

Values of the complex refractive index \( \tilde{n}_j \) of the individual minerals detected in our samples are listed in Table 4. We have restricted the calculations to two spectral domains: the near UV-visible between 370 and 700 nm and the thermal IR from 8 to 15 µm, where values for all individual minerals are available.

Results are presented in Figure 11, where they are compared to the values of the “desert” dust aerosol model in the OPAC (Optical Properties of Aerosols and Clouds) database (Hess et al., 1998), often used in satellite retrieval algorithms in the thermal infrared (see Table 1 in Klüser et al., 2012), but also in radiative transfer calculations (Highwood et al., 2003; Haywood et al., 2005). The “desert” model in the OPAC database represents the properties of mineral dust at source region, considered as a mixture of quartz and clays in different size fractions.

We restrict Figure 11 to the visible and infrared imaginary parts of the complex refractive index \( \tilde{n}_i \). In the visible, the imaginary part of the complex refractive index is relatively constant with wavelength. Mean values are of the order of 0.003, with standard deviation of the order of 0.001, with the exception of data from the PSA4 source, which average at 0.006 and present a standard deviation (0.003). It is difficult to judge on this variability as it is induced by one sample only presenting an outstanding iron oxide fraction. In the infrared, the imaginary part shows a band-type structure in correspondence to the major absorption bands of clays and quartz, mostly found in the atmospheric window between 8 and 11 µm. There is very little sample-to-sample variability in the position of the bands, as well as in the relative proportions of the band peaks, with the exception of the Sahelian dust samples, which
present a larger standard deviation due to the enhancement of the band at 9.2 \mu m corresponding to the outstanding quartz fraction of sample SOP1-8. There is very little difference in the spectra of sources PSA3, PSA4 and to a lesser extent PSA5, which have the highest absorption values at 9.6 \mu m. However, the spectral dependence for dust from source PSA2 is different (peak at 9.2 \mu m more pronounced than that at 9.6 \mu m).

The composition-based refractive indices calculated in this work are very different to those of the OPAC database. Below 0.6 \mu m, the volume-averaged values are on the lower side than the OPAC values, with the exception of values for the PSA4 samples, due to their larger standard deviation. Above, the two datasets approach. The spectral dependence is also inconsistent, and reflects the reduced spectral variability of the values for iron oxides (both hematite and goethite) in Bédidi and Cervelle (1993) with respect to other commonly used databases, such as that of Shettle (1979) for hematite on which OPAC is based.

The calculated volume-average infrared values in the absorption band between 8 and 10 \mu m are two times higher than in the OPAC database. The main reason for that is the volume-mixing rule which enhances absorption. Additionally, for single-mineral species, various authors (Hudson et al. 2008a, 2008b; Mogili et al., 2007, 2008) found strong differences in peak positions and spectral shape between extinction spectra measured by Fourier-transform infrared spectrometry and those calculated from published optical constants. All these studies show residual differences suggesting that optical constants might vary depending on whether they had been determined on powders or bulk samples, on the degree of water association of minerals, and on their chemical form (substitutions, impurities). As this variability might be even larger when looking at actual dust samples, the comparison clearly suggests that new measurements of IR extinction spectra on real dust samples representative of the mineralogical composition of the different source areas are required.

4.2.2. Implications for Fe fractional solubility

The solubility of dust iron is a function of iron speciation, i.e. oxidation states, bonding environments and mineralogy (Journet et al., 2008; Schroth et al., 2009). Thus, due to higher solubility of iron trapped in the crystal lattice of aluminosilicates and the large abundance of clay minerals, Journet et al. (2008) suggested that the use of hematite as a single source of $DFe$ in biogeochemical models might cause an underestimation of the $DFe$ supply to the ocean. On this assumption, Paris et al. (2011) showed that the iron fractional solubility from analogue of mineral dust can be estimated as the sum of Fe solubility of various minerals compounding iron in the dust matrix. We propose to apply this calculation to this dataset to estimate the solubility of dust in source region. For this purpose, we use the iron solubility
from minerals measured by Paris et al. (2011) at pH = 4.7 which a typical pH of African
rainwaters (Desboeufs et al., 2010), i.e. $\text{SFe}$ equal to 0.17% for illite and 0.006% for free iron
from goethite and hematite. Paris et al. (2011) made the approximation that all the iron
associated to clay is present as illite. Journet et al. (2008) emphasized that even the iron in
impurities into kaolinite or feldspars matrix present a high solubility. Our measurements show
the predominance of kaolinite in comparison to illite, so we consider also the dissolution of
Fe from kaolinite due to the large proportion of this mineral in the studied samples. We use a
$\text{SFe} = 0.63\%$ for kaolinite (Paris, personal communication). The percent quantity of feldspar
being usually inferior to 1% in the samples, we neglect the iron associated to these minerals.
Smectites, such as montmorillonite, are also minerals with high content of iron with a higher
solubility (Journet et al., 2008). However, due to the difficulties in quantifying the smectite
fraction, this other source of structural Fe is not well constrained. Moreover, Shi et al. (2011)
mention the presence of highly-soluble nanoparticles of ferrihydrite or poorly crystallized iron
in fine fraction of soils. However, the quantification of this iron phase is obtained by
sequential extraction and is difficult to apply on field samples due to the low mass. So we are
not able to consider this species in our calculation.

Among the samples present in our database, the mineralogical composition including
goethite, hematite, illite and kaolinite is evaluated for 27 samples, enabling to calculate the
iron fractional solubility (Table 3). In this case, the iron content in illite is assumed equal to
4.0 ± 0.9%, which corresponds to the mean stoichiometric value of two types of illites
(Fithian, Illinois and Rochester, Minnesota, as shown in Journet et al., 2008). The iron
content in illite can vary from 0.8% to 8.4 % (Murad and Wagner, 1994), this large range is
due to the fact that Fe is present as substitution of Al or K in the sheet-like structure of illite,
and this will depend on the environmental conditions that the clay has experienced in the soil
fraction (weathering,...). In goethite and hematite, we estimated the iron content based on the
analysis of Journet et al. (2008) and Paris et al., (2011) as 62.9% and 57.5% for goethite and
hematite, respectively. For kaolinite, Mestdagh et al. (1980) show that the iron content is
related to the cristallinity of the mineral and ranges from 0.02% to 0.81% with an average of
0.4% ± 0.2%, which is used for our calculation.

In order to have a larger view of the variability of iron solubility as a function of emission
source, we can also calculate iron solubility from the partitioning between structural and free
iron as obtained by the CBD extraction analysis (Table 2). This is doable for 85 samples,
among which the 27 samples for what the mineralogical composition is estimated. In this
case, we consider the structural iron ($\%\text{Fe}_{\text{structural}}$) is associated to illite ($\%\text{Fe}_{\text{illite}}$) and kaolinite
($\%\text{Fe}_{\text{kaolinite}}$) in the proportion of illite-to-kaolinite ratio corresponding to the origin of the
samples (see 3.2), such as:
\[
\%Fe_{\text{kaolinite}} = \%Fe_{\text{structural}} \left(1 + 10 \times \frac{I}{K}\right)
\]

and

\[
\%Fe_{\text{illite}} = \%Fe_{\text{structural}} - \%Fe_{\text{kaolinite}}
\]

with 10 represents the illite-to-kaolinite iron content ratio (4.0%/0.4%)

Results are summarized in Figure 12.

The mean value of iron fractional solubility is 0.10% (± 0.02%) whatever the methods of calculation, i.e. from mineralogical composition or from CBD extraction analysis. There is little variability of iron solubility as a function of the source; nonetheless, this is directly opposite to the iron oxide content, i.e., the lowest values (0.08% ± 0.01%) obtained for samples of dust emitted by local erosion in the Sahel, in agreement with the highest iron oxides content. Inversely, the samples with the highest solubility (0.13% ± 0.01%) issued from PSA3, i.e. South Algeria/Mali which present a low % of free iron (49%). Data show that 54 to 87% of soluble iron is structural iron associated to illite, meaning that this mineral is the major provider of soluble iron in our conditions of calculation. The determining factor for this parameter is the quantification of illite and hence the hypothesis made on the illite-to-kaolinite ratio as a function of emission source, which is estimated from Caquineau et al. (2002).

Thus, the regional variability on iron solubility obtained here is mainly associated with this ratio which should be better constrained. A rapid calculation for the PSA3 samples, the richest in structural iron, show that the iron solubility is divided by a factor 2 when I/K varies from 0.1 to 2.3, the extreme values observed by Caquineau et al. (2002). This means that the solubility between the various emission sources varies at the best by a factor 2. Another source of error could be the estimation of iron oxide which is affected by 30% errors (and those of illite too, by extension). Sample-to-sample comparison shows that there is a rather good correspondence with the two methods of calculation (within 15%). However, the calculation based on mineralogy yields values lower by a factor of 1.5 for the samples issued from PSA3 alone or in mixing with other sources, meaning an underestimation of structural iron in comparison from the direct measurements by CBD analyses. Smectites are identified for several samples of this source, we underestimate so the calculated solubility for the samples where these minerals are present.

The mean fractional iron solubility obtained by calculation is consistent but rather inferior to the observations obtained for transported Saharan dust over Atlantic which ranges from 0.12 to 4.1% (Sarthou et al., 2003; Baker et al., 2006; Sholkovitz et al., 2012). However, our calculation are exclusively based on mineralogical parameters, since our samples have been
collected at most 2 days after emission and do not integrate the various chemical and physical processes increasing dust solubility during long range transport, as cloud process (Desboeufs et al., 2001); photochemical process (Hand et al., 2004) or organic complexation (Paris et al., 2013). Thus, the range of calculated values could be representative of iron solubility for the various sources of African dust before transport. Moreover, the limited regional variability estimated here in comparison to the variability of measurements over Atlantic Ocean show that the emission source of dust is less critical than the process after transport to estimate the fractional iron solubility of dust. Nevertheless, new direct measurements of the iron fractional solubility for sources of dust in Western Africa are necessary to validate these conclusions.

5. Final remarks, conclusions and perspectives

In this paper we have presented an insight on the mineralogical composition of mineral dust from Western Africa from the synthesis of the airborne and ground-based observations of field campaigns conducted between 2006 and 2007. These observations have been performed at emission or within 2 days of transport in the atmosphere. The strength of this synthesis resides in the fact that the observations have been performed in a consistent way from the methodological point of view (both for the sample collection and analysis) and that they span a large geographical region covering some of the major African dust sources. The composition data presented in this paper are being made public on a dataset called DUST-MAP, to date still on its trial version. Meanwhile, data are available upon request. The perspective evolution of this dataset will be the inclusion of data from other field campaigns, both close to source regions and at some distance. Amongst those are the FENNEC and the SAMUM field campaigns (Heintzenberg et al., 2008; Ansmann et al., 2011; Washington et al., 2013), which were conducted downwind sources in Algeria and Morocco relevant to large-scale transport (Caquineau et al., 2002). This will require a close-up examination of experimental protocols and a careful evaluation of their influence on the results. The presentation of this work deserves some final considerations and opens various perspectives.

The bulk mineralogical composition of dust in Western Africa can be described by a simple model where kaolinite and illite account for 80% or above of the mass, and quartz, iron and titanium oxides for the remaining 20%. Calcite is a distinctive element of Saharan sources for which it ranges between 0.3 and 8.4% in dust mass. For the remaining minerals, the variability is due to source type (e.g., the proportions of illite and kaolinite as evidenced by Caquineau et al. (2002)) or to differences in the size distribution, particularly the increase of the quartz fraction corresponding to a coarse mode centred around 8 µm in diameter. For the
first time to our best knowledge, we provide with a large dataset of titanium oxide data which can be relevant to atmospheric photochemistry (N'Dour et al., 2009). We estimate that Sahelian dust is composed by 2% (by mass) of titanium oxides, that is, double the mass fraction of dust from sources in the PSA2, PSA3, PSA4 and PSA5 areas, with very little sample-to-sample variability. Iron oxides, also important for atmospheric chemistry processes and radiation, display higher values in Sahelian than in Saharan dust.

The discussion of the regional variability of the mineralogical composition is based on our ability in detecting source areas based on satellite remote sensing or on soil properties map (e.g., Claquin et al., 1999; Schepanski et al., 2007; Laurent et al., 2008; Ginoux et al., 2012; Nickovic et al., 2013; Journet et al., 2014), and in attributing the provenience of an air mass to it via back-trajectory calculations (e.g., Klaver et al., 2011). Beside the uncertainties inherent to the detection of sources, there are two additional factors of ambiguity in source attribution: first, air masses, especially during winter-time, mostly travel in the boundary layer and might uplift dust aerosols at various times before reaching the sampling site; secondly, aircraft samples might integrate transport from various sources active at the same time due to the fact that aircraft moves during sampling and that sample collection lasts at least 20 minutes, which corresponds to about 120 km at typical aircraft cruise speed (see Figure 1 in Chou et al., 2008 and Figure 1 in Klaver et al., 2011). It is therefore clear that some fine features of the mineralogy specific to individual sources might be lost. Nevertheless, our data well represent the composition of the regional dust load as it results from the natural mixing occurring during transport. They are also of comparable spatial resolution to that of various state-of-the-arts global and regional circulation models including dust mineralogy (Balkanski et al., 2007; Hoose et al., 2008; Huneeus et al., 2010). Henceforth, the composition data presented in this paper may serve to evaluate the regional and global model estimates of the dust mineralogical composition in the Western African atmosphere and to calculate dust properties relevant to their direct, indirect and biogeochemical impacts. Claquin et al. (1999) have discussed some comparisons between the mineralogy of dust atmospheric data and that of the soil fraction predicted by a surface mineralogy table constructed from maps of the arid soil mineralogy (updates of this table have been recently published by Nickovic et al. (2012) and Journet et al. (2014)). To do so, these authors have calculated weight ratios between some minerals (calcite, feldspar and hematite, where hematite is meant to represent the sum of the content of goethite and hematite) with respect to quartz. Our data are consistent with the Claquin et al. (1999) dataset regarding the variability of the calcite-to-quartz ratio, but display higher and lower ratios for hematite and feldspar, respectively. Our ratio of hematite-to-quartz ranges between 0.03 and 1 (0.09 and 3 when the sum of goethite and hematite is considered) instead than 0.009–0.09, whereas our ratio feldspar-to-quartz is
in the range 0.03–0.11 instead that 0.2–2 in Claquin et al. (1999). These differences illustrate
the fact that our data are closer to representing the clay soil fraction, whereas the data of
Claquin et al. (1999) best represent also the silt fraction, richer in quartz particles. When
limited to sample SOP1-8, the local erosion sample enriched in quartz and in the particle
fraction larger than 20 µm, ratios are 0.09 for hematite-to-quartz (0.03 if hematite only is
considered) and 0.08 for feldspar-to-quartz, in the range expressed by Claquin et al. (1999).

Finally, we have provided with the estimate of parameters relevant to the direct impact of
mineral dust on radiation and on ocean productivity: the complex refractive index \( \bar{n} \) and the
iron fractional solubility \( S_{Fe} \).

Assuming internal mixing, we have shown that the complex index of refraction based on
mineralogy has some variability mostly related to the partitioning between illite and kaolinite
and to the quartz fraction higher than 40% of the dust volume. The spectral behaviour of the
estimated complex index of refraction is quite different from the OPAC database. The OPAC
approach overestimates absorption in the visible. On the other hand, a very recent study
presents values of the complex refractive index in the shortwave based on a global
mineralogy soil map (Schmid et al., 2013). At 550 nm, these authors predict a real part of the
complex refractive index around 1.56, that is, higher than our which ranges between 1.45
and 1.51. Their imaginary part averaging at \( 6.2 \times 10^{-4} \), that is, almost an order of magnitude
lower than our predicted values (average \( 3 \times 10^{-3} \)). This is the consequence of the fact that
optical properties estimated from the soil fraction will reflect more closely the optical
properties of quartz, which, is enriched in the soil fraction but which in the aerosol fraction
due to the size fractionation at emission. An additional confirmation is given by the fact that,
in our data, the highest value of the real part of the complex refractive index is obtained for
sample SOP1-8, representing the Sahelian erosion case with the highest quartz fraction
(41% by weight), whereas the highest values of the imaginary part are obtained for samples
representing long-range transport from Libya through north Niger, having a low or even the
lowest quartz percentage by weight (1.5–6%) but having the highest calcite fraction (4–8%
by weight). As a matter of fact, quartz is transparent in the visible, whereas calcite, but also
clays are absorbing. These considerations bring to the conclusion that not only the soil
mineralogy but also the transfer function between the soil and the aerosol fractions must be
implemented in models estimating the dust optical properties.

In the longwave spectrum, our calculations suggest higher absorption than the OPAC model
in the atmospheric window. The 10–12 µm band, which is sensitive to differences in the
mineralogical composition amongst sources, is relevant to the retrieval of various space-
borne sensors, including IASI, AIRS, and MODIS (De Souza Machado et al., 2006; Hudson
et al., 2008; Klüser et al., 2011; Laskina et al., 2012). Some of the discrepancies between
our values and the OPAC database to be, at least partially, due to differences in the optical constants of the individual minerals. Dedicated in situ measurements of the optical constants of the dust samples would allow confirming this hypothesis.

Nonetheless, a limitation of this work, impacting our estimate of the complex refractive index, is the quantification of smectite-type clays. For the purposes of the present study, we have chosen to neglect the contribution of smectites, which were detected only on a few samples, and which we consider as a trace component of the aerosol. Also, smectites are dominant in the 0.2 µm size fraction, which is poor in mineral dust in our samples (Klaver, 2012).

In the visible Egan and Hilgeman (1979) indicated that two different montmorillonite types (one from Wyoming and one from Mississippi) have very similar refractive indices, to those of either kaolinite or illite, both in spectral dependence and magnitude. However, below 500 nm the imaginary part of the complex refractive index of the montmorillonite from Mississippi is higher by a factor of 2 than that of illite, reaching 0.004 between 200 and 360 nm. In the infrared, various authors (Toon et al., 1976; Glotch et al., 2007; Hudson et al., 2008) have shown that montmorillonite has similar optical properties to those of illite, albeit displaying higher intensity of extinction around 9.5 µm, in the Si–O stretching band. Köster et al. (1999) examined 5 types of nontronites and Fe-rich smectites and found that the position and the intensity of this band vary depending on tetrahedral Fe(III) content. Also, the speciation of clays, in particular that of montmorillonite, has implications for the cloud and ice nucleation properties of mineral dust and the estimate of iron fractional solubility (Hoose et al., 2008; Journet et al., 2008). Ultimately, these facts depend on the compositional heterogeneity in Al-rich smectite in the natural soils and sediments (Christidis and Dunham 1993, 1997; Christidis, 2006), due to various processes, including weathering and/or hydrothermal alteration of basic rocks (Köster et al. 1999), but also reduction reactions of structural Fe by microorganisms (Pentrakova et al., 2013). This makes their quantification extremely dependent on the environmental conditions. Similar considerations will held for chlorite-type clays, which were not detected in our samples, but might be present in dust from North African deserts such as Morocco and Tunisia (Paquet et al., 1984; Kandler et al., 2009). It should be noted that to date no measurements of the complex refractive index of chlorites are available in the shortwave. In the longwave, chlorite has significantly different optical properties than illite and kaolinite, in particular concerning their major absorption bands, which is located at 9.6 µm for illite and kaolinite, and at 10.2 µm for chlorite (Egan and Hilgeman, 1979; Mooney and Knacke, 1985). Should chlorite be detected in the samples, its lack of quantification will be treated as an additional source of error to be evaluated.

Regarding the iron fractional solubility, our data are not conclusive in identifying a clear regional difference with respect to the variability observed for dust collected over the Atlantic
Ocean after short- and long-range transport. This limited regional variability suggests that a single reference value (0.1%) of the fractional solubility $S_{Fe}$ of Saharan and Sahelian dust before atmospheric transport could be used in biogeochemical models. Direct field measurements of $S_{Fe}$ close to emission sources, unavailable at the moment, are needed to confirm this finding. Finally, the $S_{Fe}$ value estimated for dust near source regions is in the lowest range than values obtained for dust collected over the Atlantic Ocean, pointing to the need of taking atmospheric processing into account when iron solubility is ought to be described at the global scale.

Acknowledgements

This research was funded by several institutions. Based on a French initiative, AMMA was built by an international scientific group and is currently funded by a large number of agencies, especially from France, the United Kingdom, the United States, and Africa. It has been the beneficiary of a major financial contribution from the European Community’s Sixth Framework Research Programme. Detailed information on scientific coordination and funding is available on the AMMA International Web site at www.amma-international.org. Financial support by the API-AMMA and LEFE (project BIRD) French national funding programs is also acknowledged. The participation of P. Formenti to the DODO and GERBILS field campaigns was supported and by the UK NERC and by the UK MetOffice, which are gratefully acknowledged.

Thanks are also due to the LISA personnel who participated in field campaigns and helped with sample collection, as well as to the BAe-146 and SAFIRE air and ground crews, as well as the FAAM and Met Office observers.

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Table captions

Table 1. Calibration coefficients by mineral obtained for the calibration of the XRD analyser used in this study. The calibration coefficients represent the slope of the correlation line between the number of diffracted counts by unit mass. The references of the standard minerals used in this study are also reported.

Table 2. Mean values of the relative proportions of iron in the form of hematite and goethite to total iron oxide as obtained by XAS/XANES analysis. Standard deviations are indicated in parenthesis.

Table 3. Mineralogical composition (percentage by mass, in %) according to the source region. Standard deviations are indicated in parenthesis.

Table 4. Complex refractive indices of individual minerals used in this study. The spectral domains are indicated. *Longtin et al. (1988) have compiled values of complex refractive indices for hematite and quartz based on different sources.

Figure captions

Figure 1. Comparison between the total estimated dust mass (TEDM) estimated from the chemical composition and the total dry gravimetric mass (TDGM) for the AMMA SOP0 (open circles) and AMMA SOP1-2 data (black circles). The dashed line represents the slope of the calculated linear regression.

Figure 2. Geographical identification of the origin of the analysed samples superimposed to the localisation of the major potential source areas (PSA) of mineral dust in western and northern Africa proposed by Formenti et al. (2011a) and Scheuvens et al. (2013) on the basis of the analysis of satellite products and chemical/mineralogical composition on the aerosol and the parent soil. The approximate localisation of the Sahelian dust sources is also indicated by the shed grey area. The location of the Banizoumbou ground-based site is indicated by the black cross. The operating areas of the AMMA, DABEX, DODO and GERBILS field projects are shown. Figure reproduced from Scheuvens et al. (2013) with permission from Elsevier.

Figure 3. Scatterplot of the elemental ratio Fe/Ca versus Si/Al with additional information on source identification based on back-trajectory/satellite and dispersion modeling.

Figure 4. Range of variability of the mineral content (in percent) as obtained by XRD analysis. Boxes indicate the 25, 50 and 75% percentiles. Whiskers indicate the minimum and the maximum values, whereas the open squares indicate the mean value of the distribution.

Figure 5. Variability of the iron oxide content (ratio of Fe_{ox} to total elemental Fe) according to the source region. Boxes indicate the 25, 50 and 75% percentiles. Whiskers indicate the minimum and the maximum values, and points indicate the 5 and 95% percentiles. Data labels are as follows: Mo = Morocco; Ma = Mali; Mau = Mauritania; SA = South Algeria; NN = North Niger. Figure 6. Dependence of the Fe_{ox} content (Fe_{ox}–to–Fe ratio) on the Fe/Ca ratio for samples of different source regions identified in this work.

Figure 7. Scatterplot of the estimated mass of TiO_2 with respect to the total dust mass (TDM). Open circles are used to display values for samples corresponding to transport of Saharan dust, whereas filled circles are for samples corresponding to local emission in the Sahel.

Figure 8. Scatterplot of the Ti/Fe ratio with respect to the Fe_{ox}-to-Fe ratio.

Figure 9. Normalized volume size distribution dV/dlog(EOD) at Banizoumbou during three consequent days in summer (J1 red line, J2 green line, J3 blue line) and one day in winter (black line) for samples representing dust transported from the Bodélé source region. Data are reported as a function of the equivalent optical diameter (EOD) as obtained by optical counter measurements without any corrections for the sample refractive index.
Figure 10. Normalised volume size distribution dV/dlog(EOD) for samples representing dust of emitted dust in the Sahel during episodes local erosions. The explanation of the sample ID is reported in the text.

Figure 11. Imaginary part of the complex refractive index calculated from the mineralogical composition of dust originating from sources PSA5, PSA4, PSA3, local erosion in the Sahel, and PSA2 (from top to bottom). Figure 11.a represents the real part in the 0.3-0.7 µm, calculated assuming refractive index from Table 3. Grey shaded curves represent results obtained when using the data from Bedidi and Cervelle (1993) for hematite, whereas blue shaded areas correspond to using values from Shettle (1979). Figure 11.b represents the imaginary part of the complex refractive index calculated in the 6-16 µm spectral domain. The black line with white diamonds represents data from the OPAC database (Hess et al., 1998).

Figure 12. Iron solubility calculated from the mineralogical composition and from CBD analysis as a function of emission source. Error bars indicate standard deviation. When error bar is not presented, the data is only for one sample.
Table 1. Calibration coefficients by mineral obtained from the calibration of the XRD analyser used in this study. The calibration coefficients represent the slope of the linear correlation between the number of diffracted counts by unit mass. The references of the standard minerals used in this study are also reported. The absolute uncertainty on the slope is also indicated, whereas numbers in parenthesis are the percent uncertainty obtained as ratio between the absolute uncertainty and slope value.

<table>
<thead>
<tr>
<th>Mineral</th>
<th>Origin</th>
<th>Slope (cps mg(^{-1}))</th>
<th>R(^2)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Quartz (SiO(_2))</td>
<td>Fontainebleau, France</td>
<td>446 ± 14 (3%)</td>
<td>0.88</td>
</tr>
<tr>
<td>Calcite (CaCO(_3))</td>
<td>Bédarieux, France</td>
<td>325 ± 11 (3%)</td>
<td>0.94</td>
</tr>
<tr>
<td>Dolomite (CaMg(CO(_3))(_2))</td>
<td>Traversella, Italie</td>
<td>679 ± 70 (10%)</td>
<td>0.55</td>
</tr>
<tr>
<td>Gypsum (CaSO(_4) 2H(_2)O)</td>
<td>Unknown</td>
<td>446 ± 22 (5%)</td>
<td>0.89</td>
</tr>
<tr>
<td>Ortoclase (KAISi(_3)O(_8))</td>
<td>Madagascar</td>
<td>997 ± 70 (7%)</td>
<td>0.69</td>
</tr>
<tr>
<td>Albite (NaAlSi(_3)O(_8))</td>
<td>Ontario, Canada</td>
<td>2456 ± 56 (2%)</td>
<td>0.96</td>
</tr>
</tbody>
</table>
Table 2. Mean percent values of the oxide fraction of total Fe and relative proportions of hematite and goethite to total iron oxide as obtained by XAS/XANES analysis. Standard deviations are indicated in parenthesis.

<table>
<thead>
<tr>
<th>Origin</th>
<th>Oxide fraction of total Fe (%)</th>
<th>Goethite fraction of Fe oxide (%)</th>
<th>Hematite fraction of Fe oxide (%)</th>
</tr>
</thead>
<tbody>
<tr>
<td>PSA5 (Bodélé)</td>
<td>43 (± 11)</td>
<td>65 (± 7)</td>
<td>35 (± 7)</td>
</tr>
<tr>
<td>PSA3 (North Niger)</td>
<td>57 (± 2)</td>
<td>59 (± 8)</td>
<td>41 (± 8)</td>
</tr>
<tr>
<td>PSA3 (South Algeria)</td>
<td>55 (---)</td>
<td>62 (---)</td>
<td>38 (---)</td>
</tr>
<tr>
<td>Erosion, Sahel</td>
<td>66 (± 5)</td>
<td>68 (± 5)</td>
<td>32 (± 5)</td>
</tr>
<tr>
<td>Source</td>
<td>Kaolinite</td>
<td>Illite</td>
<td>Quartz</td>
</tr>
<tr>
<td>---------------------------------------------</td>
<td>-----------</td>
<td>--------</td>
<td>--------</td>
</tr>
<tr>
<td>Bodélé (wintertime)</td>
<td>84.1 (--)</td>
<td>8.5 (--)</td>
<td>2.9 (--)</td>
</tr>
<tr>
<td>Bodélé/Sudan (summertime)</td>
<td>76.1 (4.4)</td>
<td>7.0 (0.4)</td>
<td>11.2 (5.6)</td>
</tr>
<tr>
<td>North Niger/Libya</td>
<td>54.0 (--)</td>
<td>37.8 (--)</td>
<td>4.6 (--)</td>
</tr>
<tr>
<td>North Niger/Libya/south Algeria</td>
<td>60.5 (4.7)</td>
<td>24.0 (1.9)</td>
<td>3.6 (3.1)</td>
</tr>
<tr>
<td>South Algeria/Mali</td>
<td>71.1 (13.5)</td>
<td>16.4 (11.8)</td>
<td>7.1 (0.8)</td>
</tr>
<tr>
<td>South Algeria/North Niger (smectites)</td>
<td>75.2 (3)</td>
<td>7.0 (0.3)</td>
<td>10.8 (2.0)</td>
</tr>
<tr>
<td>South Algeria/North Niger (non smectites)</td>
<td>78.2 (2.2)</td>
<td>7.2 (0.2)</td>
<td>6.6 (0.7)</td>
</tr>
<tr>
<td>North Niger</td>
<td>78.3 (--)</td>
<td>7.4 (--)</td>
<td>9.7 (--)</td>
</tr>
<tr>
<td>Local erosion (all)</td>
<td>73.0 (13)</td>
<td>7.1 (1.3)</td>
<td>14.3 (13)</td>
</tr>
<tr>
<td>Local erosion (quartz-rich)</td>
<td>46.6 (--)</td>
<td>4.4 (--)</td>
<td>40.6 (--)</td>
</tr>
<tr>
<td>Local erosion (excluding quartz-rich)</td>
<td>78.3 (1.5)</td>
<td>7.5 (0.2)</td>
<td>9.1 (2.2)</td>
</tr>
<tr>
<td>Mali/Mauritania</td>
<td>29.7 (--)</td>
<td>46.3 (--)</td>
<td>17.3 (--)</td>
</tr>
<tr>
<td>Mali/Mauritania/Western Sahara</td>
<td>30.7 (3.2)</td>
<td>54.2 (6.0)</td>
<td>8.4 (1.7)</td>
</tr>
</tbody>
</table>
Table 4. Complex refractive indices of individual minerals used in this study. The spectral domains are indicated. *Longtin et al. (1988) have compiled values of complex refractive indices for hematite and quartz based on different sources.

<table>
<thead>
<tr>
<th>Mineral</th>
<th>Reference</th>
<th>Spectral domain (µm)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Illite</td>
<td>Egan et Hilgeman (1979)</td>
<td>0.185-2.6</td>
</tr>
<tr>
<td></td>
<td>Querry (1987)</td>
<td>2.5-200</td>
</tr>
<tr>
<td></td>
<td>Glotch et al. (2007)</td>
<td>5-100</td>
</tr>
<tr>
<td>Kaolinite</td>
<td>Egan et Hilgeman (1979)</td>
<td>0.185-2.6</td>
</tr>
<tr>
<td></td>
<td>Roush et al. (1991)</td>
<td>5-25</td>
</tr>
<tr>
<td></td>
<td>Glotch et al. (2007)</td>
<td>5-100</td>
</tr>
<tr>
<td>Calcite</td>
<td>Querry (1978)</td>
<td>0.2-25</td>
</tr>
<tr>
<td></td>
<td>Long et al. (1993)</td>
<td>2.5-300</td>
</tr>
<tr>
<td>Dolomite</td>
<td>Barthelmy (2007)</td>
<td>0.185-2.5</td>
</tr>
<tr>
<td></td>
<td>Querry (1987)</td>
<td>2.5-40</td>
</tr>
<tr>
<td>Albite</td>
<td>Barthelmy (2007)</td>
<td>0.185-2.5</td>
</tr>
<tr>
<td>Ortoclase</td>
<td>Barthelmy (2007)</td>
<td>0.185-2.5</td>
</tr>
<tr>
<td>Gypsum</td>
<td>Barthelmy (2007)</td>
<td>0.185-2.5</td>
</tr>
<tr>
<td></td>
<td>Long et al. (1993)</td>
<td>2.5-300</td>
</tr>
<tr>
<td>Goethite</td>
<td>Bedidi et Cervelle (1993)</td>
<td>0.46-0.7</td>
</tr>
<tr>
<td></td>
<td>Glotch et Rossman (2009)</td>
<td>8.3-50</td>
</tr>
<tr>
<td>Hematite</td>
<td>Bedidi et Cervelle (1993)</td>
<td>0.4-0.7</td>
</tr>
<tr>
<td></td>
<td>Marra et al., (2005)</td>
<td>6.5-50</td>
</tr>
<tr>
<td></td>
<td>Longtin et al. (1988)*</td>
<td>0.2-300</td>
</tr>
<tr>
<td>Quartz</td>
<td>Longtin et al.,(1988)*</td>
<td>0.2-2</td>
</tr>
<tr>
<td></td>
<td>Peterson et Weinman (1969)</td>
<td>5-36</td>
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<td>Spitzer et Kleinman (1961)</td>
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Figure 1. Comparison between the total estimated dust mass (TEDM) estimated from the chemical composition and the total dry gravimetric mass (TDGM) for the AMMA SOP0 (open circles) and AMMA SOP1-2 data (black circles). The dashed line represents the slope of the calculated linear regression.

\[
y = 0.98x
\]

\[
R^2 = 0.88
\]
Figure 2. Geographical identification of the origin of the analysed samples superimposed to the localisation of the major potential source areas (PSA) of mineral dust in western and northern Africa proposed by Formenti et al. (2011a) and Scheuvens et al. (2013) on the basis of the analysis of satellite products and chemical/mineralogical composition on the aerosol and the parent soil (Claquin et al., 1999; Brooks and Legrand, 2000; Caquineau et al., 2002; Prospero et al., 2002; Washington et al., 2003; Schepanski et al., 2007; 2009; 2012; Laurent et al., 2008; Klüser and Schepanski, 2009). The approximate localisation of the Sahelian dust sources is also indicated by the shed grey area. The location of the Banizoumbou ground-based site is indicated by the black cross. The operating areas of the AMMA, DABEX, DOD and GERBILS field projects are shown. Figure reproduced from Scheuvens et al. (2013) with permission from Elsevier.
Figure 3. Scatterplot of the elemental ratio Fe/Ca versus Si/Al with additional information on source identification based on back-trajectory/satellite and dispersion modeling.
Figure 4. Range of variability of the mineral mass fraction (in percent) as obtained by XRD analysis. Boxes indicate the 25, 50 and 75% percentiles. Whiskers indicate the minimum and the maximum values, whereas the open squares indicate the mean value of the distribution.
Figure 5. Variability of the iron oxide content (ratio of Fe_{ox} to total elemental Fe) according to the source region. Boxes indicate the 25, 50 and 75% percentiles. Points indicate the minimum and the maximum values, and Whiskers indicate the 5 and 95% percentiles. Data labels are as follows: Mo = Morocco; Ma = Mali; Mau = Mauritania; SA = South Algeria; NN = North Niger.
Figure 6. Dependence of the Fe$_{\text{ox}}$ content (Fe$_{\text{ox}}$–to–Fe ratio) on the Fe/Ca ratio for samples of different source regions identified in this work.
Figure 7. Scatterplot of the estimated mass of TiO$_2$ with respect to the total dust mass (TDM). Open circles are used to display values for samples corresponding to transport of Saharan dust, whereas filled circles are for samples corresponding to local emission in the Sahel.
Figure 8. Scatterplot of the Ti/Fe ratio with respect to the Fe$_{ox}$/Fe ratio. Open circles are used to display values for samples corresponding to transport of Saharan dust, whereas filled circles are for samples corresponding to local emission in the Sahel. Two grey points indicate identified mixtures.
Figure 9. Normalized volume size distribution $dV/d\log(\text{EOD})$ at Banizoumbou during three consequent days in summer (J1 red line, J2 green line, J3 blue line) and one day in winter (black line) for samples representing dust transported from the Bodélé source region. Data are reported as a function of the equivalent optical diameter (EOD) as obtained by optical counter measurements without any corrections for the sample refractive index.
Figure 10. Normalised volume size distribution $dV/d\log(EOD)$ for samples representing Sahelian dust: SOP1-8 representing an episode of local emission by erosion; SOP1-15 and SOP1-60 representing the background dust composition in the absence of local erosion; and SOP1-17 representing the composition of Sahelian dust advected at the sampling site after having been emitted by local erosion in the proximity. Numbers represent the measured percent mass fraction of quartz and feldspars (sum of albite and orthoclase), respectively.
Figure 11. Imaginary part of the complex refractive index calculated from the mineralogical composition of dust steaming from sources PSA5, PSA4, PSA3, local erosion in the Sahel, and PSA2 (from top to bottom). Figure 11.a represents the real part in the 0.3-0.7 µm, calculated assuming refractive index from Table 3. Grey shaded curves represent the mean values (light grey line) and 1 standard deviation (grey shaded areas) obtained when using the data from Bedidi and Cervelle (1993) for hematite, whereas blue shaded areas correspond to values calculated using the hematite data from Shettle (1979). The light blue line represents mean values and the blue shaded area represents one standard deviation. Figure 11.b represents the imaginary part of the complex refractive index calculated in the 6-16 µm spectral domain. The black line with white diamonds represents data from the OPAC database (Hess et al., 1998).
Figure 11. Imaginary part of complex refractive index vs. Wavelength, µm for different samples: OPAC, PSA5, PSA4, PSA3, Sahelian dust, PSA2.
Figure 12. Iron solubility calculated from the mineralogical composition and from CBD analysis as a function of emission source. Error bars indicate standard deviation. When error bar is not presented, the data is only for one sample.