Climatology of aerosol component concentrations derived from multi-angular polarimetric POLDER-3 observations using GRASP algorithm

Lei Li1, Yevgeny Derimian2, Cheng Chen2,3, Xindan Zhang1, Huizheng Che1, Gregory L. Schuster4, David Fuertes3, Pavel Litvinov3, Tatyana Lapyonok2, Anton Lopatin3, Christian Matar3, Fabrice Ducos2, Yana Karol3, Benjamin Torres2, Ke Gui1, Yu Zheng1, Yuanxin Liang1, Yadong Lei1, Jibiao Zhu1, Lei Zhang1, Junting Zhong1, Xiaoye Zhang1, and Oleg Dubovik2

1State Key Laboratory of Severe Weather & Key Laboratory of Atmospheric Chemistry of CMA, Chinese Academy of Meteorological Sciences, Beijing 100081, China
2Univ. Lille, CNRS, UMR 8518 – LOA – Laboratoire d’Optique Atmosphérique, 59000 Lille, France
3GRASP-SAS, Villeneuve d’Ascq, France
4NASA Langley Research Center, Hampton, VA, USA

Correspondence: Huizheng Che (chehz@cma.gov.cn) and Oleg Dubovik (oleg.dubovik@univ-lille.fr)

Abstract. The study presents a climatology of aerosol composition concentrations obtained by a recently developed algorithm approach, namely the Generalized Retrieval of Atmosphere and Surface Properties (GRASP)/Component. It is applied to the whole archive of observations from the POLarization and Directionality of the Earth’s Reflectances (POLDER-3). The conceptual specifics of the GRASP/Component approach is in the direct retrieval of aerosol speciation (component fraction) without intermediate retrievals of aerosol optical characteristics. Although a global validation of the derived aerosol component product is challenging, the results obtained are in line with general knowledge about aerosol types in different regions. In addition, we compare the GRASP-derived black carbon (BC) and dust components with those of the Modern-Era Retrospective Analysis for Research and Applications, version 2 (MERRA-2) product. Quite a reasonable general agreement was found between the spatial and temporal distribution of the species provided by GRASP and MERRA-2. The differences, however, appeared in regions known for strong biomass burning and dust emissions; the reasons for the discrepancies are discussed. The other derived components, such as concentrations of absorbing (BC, brown carbon (BrC), iron-oxide content in mineral dust) and scattering (ammonium sulfate and nitrate, organic carbon, non-absorbing dust) aerosols, represent scarce but imperative information for validation and potential adjustment of chemical transport models. The aerosol optical properties (e.g., aerosol optical depth (AOD), Ångström exponent (AE), single-scattering albedo (SSA), fine- and coarse-mode aerosol optical depth (AODF AND AODC)) derived from GRASP/Component were found to agree well with the Aerosol Robotic Network (AERONET) ground reference data, and were fully consistent with the previous GRASP Optimized, High Precision (HP) and Models retrieval versions applied to POLDER-3 data. Thus, the presented extensive climatology product provides an opportunity for understanding variabilities and trends in global and regional distributions of aerosol species. The climatology of the aerosol components obtained in addition to the aerosol optical properties provides additional valuable, qualitatively new insight about aerosol distributions and, therefore, demonstrates advantages of multi-angular polarimetric (MAP) satellite observations as the next frontier for aerosol inversion from advanced satellite observations. The extensive satellite-based aerosol component dataset is expected to be useful for improving global aerosol emissions and component-resolved radiative forcing estimations. The GRASP/Component products are publicly available (https://www.grasp-open.com/products/, last access: 15 March 2022) and the dataset used in the current study is registered under https://doi.org/10.5281/zenodo.6395384 (Li et al., 2022b).
1 Introduction

The latest Intergovernmental Panel on Climate Change (IPCC) reported that aerosols are still the most dominant contributor to the large forcing uncertainty (Forster et al., 2021), with the assessment of $-2.0$ and $-0.4 \text{ W m}^{-2}$ (90\% likelihood) for the aerosol effective radiative forcing (Bellouin et al., 2020b). One of the main factors responsible for the large aerosol radiative forcing uncertainty is the lack of understanding and knowledge regarding global aerosol composition distribution. The potential of using remote-sensing observations has already demonstrated this in several studies. For example, Dubovik et al. (2002b) discussed the differences of spectral absorption between carbonaceous particles and mineral dust using retrieval from the global Aerosol Robotic Network (AERONET) measurements. Schuster et al. (2005), for the first time, demonstrated a potential to infer the black carbon (BC) content from remote-sensing AERONET data using an assumption that all aerosol absorption could be interpreted to the contribution of the BC component. Then, the hematite and brown carbon (BrC) proportions were successfully estimated by Koven and Fung (2006) and Arola et al. (2011), respectively, from AERONET products using the spectral variability of complex refraction indices. Derimian et al. (2008) discussed the difficulties in separating carbonaceous species absorption from mineral dust (largely iron oxides) absorption, and highlighted that the importance of sufficient spectral resolution of measurements on capturing the differences in absorption spectral dependence among BC, BrC, and iron oxides is essential for the determination of BC, BrC, and iron-oxide proportions. With additional constraints on spectral single-scattering albedo (SSA) and the complex refractive index (Arola et al., 2011; Koven and Fung, 2006; Schuster et al., 2005), the proportions of BC, BrC and dust can be estimated simultaneously based on AERONET products (Wang et al., 2013). Recently, Schuster et al. (2016) successfully retrieved the proportions of aerosol absorbing components (BC, BrC, and iron oxides) in fine- and coarse-mode particles of biomass burning and dust aerosols over the globe, based on the aerosol products (aerosol size distributions and complex refraction index) provided in the AERONET retrievals described by Dubovik and King (2000) and Dubovik et al. (2002b, 2006). The estimation of aerosol compositional content based on AERONET optical retrievals is also discussed in several other studies (Bahadur et al., 2012; Cazorla et al., 2013; Chung et al., 2012; Costabile et al., 2013; Dey et al., 2006; Li et al., 2013, 2015; Russell et al., 2010; Xie et al., 2014, 2017; Zhang et al., 2018). AERONET products (such as particle size distribution, wavelength dependence of extinction, SSA) are also used to distinguish the aerosol types at several regional sites (Giles et al., 2012; Logan et al., 2013). The AERONET-derived aerosol properties together with the active measurements of lidar ratio and spectral depolarization ratio were also used by Burton et al. (2012, 2014) for the classification of aerosol types from the airborne High Spectral Resolution Lidar (HSRL) measurements. Ganguly et al. (2009a, b) retrieved the concentration of several aerosol components from the combination of AERONET sun-photometer and lidar measurements. It is noted that the evaluation of aerosol components by all above approaches are conducted using the remote-sensing measurements at several discrete sites, even for the globally distributed ground-based AERONET sites.

Satellite remote sensing was extensively used in diverse efforts for improving global aerosol monitoring: for characterizing spatial aerosol distribution (Kaufman et al., 2002; Remer et al., 2008), data assimilation (Dubovik et al., 2008; Chen et al., 2018, 2019; J. Zhang et al., 2008; Shi et al., 2013), and modeling of aerosol climate effects (Bellouin et al., 2005; Jia et al., 2021; Myhre, 2009; Yu et al., 2006). A variety of inversion algorithms have been proposed and applied to different spaceborne instruments for further understanding of the atmospheric aerosol properties on large spatial scale (Bréon et al., 2011; Dubovik et al., 2019; King et al., 1999; Li et al., 2009). Specifically, a new cloud- and haze-detection algorithm was proposed for the geostationary satellite Himawari-8 Advanced Himawari Imager (AHI) measurements (Shang et al., 2017, 2019), which had been used for the analysis of diurnal variations of haze weather and the impact of aerosol optical depth (AOD) on the estimation of cloud properties and surface short-wave radiation (Letu et al., 2019, 2022). Sometimes different algorithms are proposed for the same spaceborne instruments. For example, Dark Target (DT) (Levy et al., 2013; Remer et al., 2005, 2020), Deep Blue (DB) (Hsu et al., 2006, 2013), and Multi-Angle Implementation of Atmospheric Correction (MAIAC) (Lyapustin et al., 2018) are proposed for inverting the Moderate Resolution Imaging Spectroradiometer (MODIS) observations. All these algorithms enforced the MODIS retrieval and made the MODIS aerosol database a reference aerosol product in the community. At the same time, MODIS is a single-view instrument and generally does not provide sufficient information to identify aerosol types. In contrast, the multi-angular intensity observations by the Multi-angle Imaging Spectro Radiometer (MISR) have been used successfully to identify aerosol types in several regional-scale studies, including wildfire smoke (Chen et al., 2008; Guo et al., 2013), desert dust (Guo et al., 2013; Kahn et al., 2009), and volcanic ash (Kahn and Limbacher, 2012; Scollo et al., 2012; Flower and Kahn, 2017, 2018, 2020, 2021). Nonetheless, although the aerosol classification from MISR datasets has been supported by abovementioned studies, these studies mainly provided qualitative flagging of aerosol types for observed aerosol events. In addition, because of the aerosol types defined by aerosol parameters in the look-up table (LUT), the best fit to the observations for intermediate retrievals were obtained us-
ing a discrete set of possible solutions. In this regard, the numerous studies suggested that multi-angle polarization measurements have very high information content and are most appropriate for providing potentially enhanced insight about aerosol microphysics and composition for advanced aerosol products (Hasekamp and Landgraf, 2007; Knobelspiesse et al., 2011; Mishchenko and Geogdzhayev, 2007; Mishchenko and Travis, 1997; Tanré et al., 2011). However, due to the rather limited amount of available multi-angle polarization observations and the complexity in their interpretation, the added value of satellite polarimetry on aerosol monitoring remains questionable. Indeed, polarimetry has enhanced sensitivities to numerous atmospheric parameters, and inversion algorithms are required to consider all these sensitivities adequately. Partially due to this complexity, the practical advantages of multi-angular polarimetric (MAP) retrieval were not convincingly exhibited by the available operational MAP products in the past for a long time, and only recently the advanced aerosol products (including present study) make the advantages of the polarimetry for aerosol remote sensing more evident (e.g., see discussion in Dubovik et al., 2019, 2021b). Nonetheless, as discussed by Dubovik et al. (2019, 2021b) there is a large number of planned MAP missions, and significant progress has been achieved in the development of new generation algorithms for interpretation polarimetry observation. Therefore, the impact of polarimetric data on overall aerosol science is expected to increase rapidly in coming years.

The PoLarization and Directionality of the Earth’s Reflectances (POLDER) instrument was a satellite polarimeter designed for collecting the spectral directional polarized solar radiation measurements (Deschamps et al., 1994). POLDER-3, which was launched onboard PARASOL (Polarization & Anisotropy of Reflectances for Atmospheric Sciences coupled with Observations from a Lidar), provided the longest records of multi-angle polarized measurements from March 2005 to October 2013. In order to invert the POLDER multi-angle polarized measurements, several algorithms following the idea of a conventional MODIS-like LUT approach (Kaufman et al., 1997; Tanré et al., 1997) were originally proposed to retrieve the AOD products over the ocean (Deuzé et al., 2001; Goloub et al., 1999; Herman et al., 2005) and fine-mode AOD (AODF) over land (Deuzé et al., 2001; Herman et al., 2005). However, the exploration of the full extent of the aerosol information contained in MAP measurements, became feasible only relatively recently with the development of the so-called satellite “algorithms of new generation”, such as the SRON (Netherlands Institute for Space Research) (Fu and Hasekamp, 2018; Hasekamp et al., 2011) and GRASP (Generalized Retrieval of Atmosphere and Surface Properties) (Dubovik et al., 2011, 2021a) retrievals. These algorithms consider a continuous space of aerosol microphysical properties (size distribution, refractive index), instead of using standard aerosol models, and retrieve the parameters of underlying surface simultaneously with aerosol properties to properly account for land or ocean reflection. Specifically, the data analyzed in this paper were generated in frame of the algorithm named GRASP, that was initially developed for enhanced retrieval of aerosol properties from POLDER observations (Dubovik et al., 2011). The applicability of GRASP then extended to a wide variety of diverse remote-sensing applications including both passive and active observations (see details in the papers by Dubovik et al., 2014, 2021a). After about a decade of development and advancement, the GRASP algorithm has been adapted from POLDER observations for operational generation-extended aerosol products. Several datasets of enhanced aerosol retrieval from POLDER observations (see Sect. 4 in Dubovik et al., 2021a) are now publicly available and archived at the AERONET/ICARE Data and Services Center (http://www.icare.univ-lille1.fr, last access: 20 July 2022) and GRASP-OPEN site (https://www.grasp-open.com, last access: 20 July 2022). All POLDER/GRASP retrieval releases include extended sets of conventional aerosol optical parameters (such as spectral AOD, spectral aerosol absorption optical depth (AAOD), spectral AODF, spectral coarse-mode aerosol optical depth (AODC), particle size distribution, SSA, complex refractive index, fraction of spherical particles, etc.) (Chen et al., 2020). In addition, the latest POLDER-3/GRASP-Component processing also provided some information on aerosol composition by estimation of fractions of such aerosol components as BC, BrC, absorbing and scattering mineral dust, etc. (Li et al., 2019).

This GRASP/Component approach derives fractions of aerosol components together with size distribution and non-spherical fraction of aerosol particle directly from the measured radiances without an intermediate step of optical aerosol properties’ retrieval (Li et al., 2019). This is a significant methodological difference with other approaches for aerosol typing from satellite observations. For example, Russell et al. (2014) also demonstrated a possibility of identifying aerosol types from POLDER-3 observations. However, that study analyzed aerosol optical properties derived from POLDER-3 using the SRON algorithm for aerosol classes/types (pure dust, polluted dust, biomass burning, urban/industrial, and marine aerosol), a set of criteria formulated based on the AERONET retrievals (e.g., aerosol optical and microphysical parameters). It should also be noted that the GRASP/Component approach was applied to AERONET data and was demonstrated to derive valuable information about aerosol components, as well as to provide optical properties with the accuracy comparable to the standard AERONET retrievals (Li et al., 2019; Zhang et al., 2022).

Because the validation of aerosol optical properties (AOD, Ångström exponent (AE), AODF, AODC, AAOD, SSA, etc.) generated by the POLDER-3/GRASP-Component approach has been realized and discussed by Zhang et al. (2021). This study presents the extensive analysis of aerosol composition products generated by GRASP/Component from full archive of the MAP POLDER-3 observations. We present the cli-
matological, global, statistical analysis of satellite-retrieved aerosol absorbing and scattering components with the assessment of consistency with general expectation to the extent possible. We also make comparisons of GRASP/Component BC and dust retrievals to those in Modern-Era Retrospective Analysis for Research and Applications, version 2 (MERRA-2) products. Based on preliminary sparse analysis of derived aerosol component information (Li et al., 2019, 2020a, b) and results of the study by Zhang et al. (2021) demonstrating that the component approach can provide comparable and sometimes even better aerosol optical products, our expectation is that the GRASP/Component approach can generally provide both conventional aerosol optical parameters and the parameters characterizing aerosol composition with sufficiently high accuracy from POLDER-like polarimetric data.

2 Data and methodology description

2.1 POLDER-3 sensor

The PARASOL was launched on 18 December 2004 as part of the A-Train constellation. POLDER-3, designed for the measurements of multispectral, multidirectional, and polarized radiances on board the PARASOL satellite, provides data for the period of March 2005–October 2013 (Tanré et al., 2011). POLDER-3 has nine channels (440, 490, 565, 670, 763, 765, 865, 910, and 1020 nm) for the measurements of total radiance and three channels (490, 670, and 865 nm) for the measurements of polarization. The measurements are collected in up to 16 viewing angles per pixel with a resolution of $5.3 \times 6.2 \text{ km}^2$ at nadir. Additionally, it has been demonstrated that the multi-polarization and multidirectional measurements of POLDER help to distinguish better cloudy and clear pixels (Buriez et al., 1997; Goloub et al., 1997; Parol et al., 1999; Zeng et al., 2011).

2.2 GRASP/Component product

We are discussing the advanced aerosol products obtained from POLDER-3 using the GRASP algorithm. The efficiency and flexibility of GRASP have been illustrated in many previous studies on applications for diverse remote-sensing instruments, such as ground-based AERONET photometers and lidars (e.g., Benavent-Oltra et al., 2019; Lopatin et al., 2013, 2021; Titos et al., 2019; Tsekeri et al., 2017; Zhang et al., 2022), POLDER-3 (e.g., Chen et al., 2020; Dubovik et al., 2011; Li et al., 2019, 2020a, 2020b), DPC/GF-5 (Li et al., 2022a), sky cameras (e.g., Román et al., 2017, 2022), and polar-nephelometer data (e.g., Espinosa et al., 2017; Schuster et al., 2019). The papers by Dubovik et al. (2011, 2014, 2021a) present the detailed description of the main concept of GRASP as well as the specific methodological introductions. In brief, GRASP employs the multi-term least-square minimization (Dubovik, 2004; Dubovik et al., 2021a) for the implementation of statistically optimized fitting. The advantages of this concept have been initially demonstrated in the retrievals of the aerosol properties from ground-based AERONET sun/sky-radiometers (Dubovik et al., 2000, 2002a, 2002b, 2006; Dubovik and King, 2000). One of the essential differences from traditional LUT algorithms is that GRASP performs online radiative transfer computations for multiple interactions of the scattered solar light in the atmosphere. That is, GRASP searches the optimized fitting in the continuous space of solutions, whereas the traditional LUT approaches search among the pre-computed solutions. This allows GRASP to explore a much larger parameter space (i.e., non-lognormal size distributions and orders of magnitude refractive indices) than traditional approaches. The properties of surface reflectance in GRASP are modeled using the Bidirectional Reflectance Distribution Function (BRDF) and Bidirectional Polarization Distribution Function (BPDF) models. Specifically, over land, the Ross–Li BRDF (Li and Strahler, 1992; Ross, 1981) and BPDF (Maignan et al., 2009) models are used considering geometrical and volumetric terms of the Ross–Li BRDF models are nearly spectrally constant (Litvinov et al., 2011). The reflective properties of ocean surface are taken into account using the Cox–Munk model (Cox and Munk, 1954) analogously to the many conventional approaches. The details are provided by Dubovik et al. (2011) and Frouin et al. (2019). The GRASP is an open-source code that is provided together with a described documentation on the GRASP-OPEN website https://www.grasp-open.com (last access: 30 November 2021).

Several different versions of aerosol optical products derived from previous GRASP (Optimized, High Precision (HP), and Models) approaches for the full archive of POLDER-3 observations have been generated and released for open access (Dubovik et al., 2019; Chen et al., 2020). Here, we provide only a brief description of these three approaches and their main differences that are relevant to the present analysis. As described in Dubovik et al. (2011), the solution vector in each of these three GRASP configurations can be written as Eq. (1)

$$a = (a_v, a_r, a_k, a_{sp}, a_{VC}, a_l, a_{brdf1}, a_{brdf2}, a_{brdf3}, a_{bpdf})^T,$$

where the elements representing the normalized logarithms of $D\langle r \rangle / \ln r$, real part of complex refractive index, imaginary part of complex refractive index, spherical particles fraction, total volume concentration, mean altitude of the aerosol layer, the parameters in BRDF and BPDF model are characterized by the vectors of $a_v, a_r, a_k, a_{sp}, a_{VC}, a_l, a_{brdf1}, a_{brdf2}, a_{brdf3}, a_{bpdf}$, respectively.

The GRASP aerosol optical products derived from Optimized, HP, and Models approaches are generated and released publicly as daily, monthly, seasonal, yearly, and climatological datasets in Level-2 and Level-3 (Chen et al., 2020). Level-2 represents the quality-filtered products with an initial resolution of satellite observations. There are two different spatial resolutions of 0.1 and 1° for the Level-3
products. Hereafter all the GRASP/Component climatology is presented at 0.1° × 0.1° spatial resolution, however, the comparisons with MERRA-2 will be at 0.5° × 0.625° (i.e., the GRASP/Component products are regressed into the same resolution of MERRA-2 at 0.5° × 0.625°). Level-0 represents the raw data and Level-1 represents daily files of retrievals considered as results of intermediate processing, which could be provided under request.

As mentioned above, GRASP is a versatile algorithm that can be applied to diverse observations and can serve as an accessible community tool for testing new and innovative ideas for interpretation of remote-sensing measurements, e.g., using synergy of different observations (Dubovik et al., 2021a; Lopatin et al., 2021). The most recent GRASP/Component approach described in the study of Li et al. (2019), can derive some size-resolved aerosol composition information attempting to discriminate the contributions of such species as BC, BrC, coarse-mode absorbing (mainly representing iron oxides in mineral dust), fine- and coarse-mode non-absorbing soluble (NAS) and insoluble components, etc. Specifically, the forward model of GRASP/Component uses the fixed refractive index for each component (see Table S1 in the Supplement), and based on the given component fractions, calculates the complex refraction indices of aerosol mixture for simulating the observations. Correspondingly, the state vector in the GRASP/Component configuration includes the fractions of components instead of refractive indices (Li et al., 2019):

\[
a = (a_v, a_{\text{frac}}, a_{\text{sph}}, a_{V_{\text{C}}}, a_h, a_{\text{brdf1}}, a_{\text{brdf2}}, a_{\text{brdf3}}, a_{\text{bpdf}})^T. \tag{2}
\]

where \(a_{\text{frac}}\) represents the aerosol component fractions.

Thus, the main conceptual difference of GRASP/Component from GRASP Optimized and HP is the retrieval of volume fractions of six components (BC, BrC, fine- and coarse-mode non-absorbing insoluble (FNAL and CNAI), coarse-mode absorbing insoluble (CAI), mainly representing iron oxides in mineral dust, and relative humidity for the host calculation) instead of direct retrieval of the real and imaginary parts of the complex refractive index at each wavelength (12 parameters in GRASP Optimized and HP). Thus, the number of unknowns is reduced in GRASP/Component compared to GRASP Optimized and HP. In addition, the spectral refractive index of each species assumed in the component model presents extra constraints on the spectral dependences of the complex refraction index of aerosol mixture. The advantage of GRASP/Component, associated with the reduction of unknowns and additional constraints on spectral dependences, was discussed by the study of Zhang et al. (2021). In the contrast, the GRASP/Models approach retrieves an even smaller number of parameters since it considers aerosol as an external mixture of aerosol components with a priori fixed all optical properties (particle size, refractive index and fraction of spheres). However, GRASP/Models seems to have some limitations in deriving such detailed parameters as AE, AODC, AODF, etc. (Chen et al., 2020).

The GRASP/Component approach assumes internal mixing of aerosol components for fine- and coarse-mode aerosols, which requires a mixing rule on the estimation of components for the aerosol mixture. Maxwell–Garnett’s (MG) effective medium approximation is employed in GRASP/Component, as one of main mixing rules, to estimate an effective refractive index of aerosol mixture considering several insoluble components presenting soluble host, such as ammonium nitrate (or sulfate) (Bohren and Huffman, 1983; Lesins et al., 2002). The hygroscopicity of ammonium nitrate has been quantitatively described in detail (Tang, 1996), thus the proportions and properties of ammonium nitrate and aerosol water content (AWC) are selected for the calculation of optical properties of the host. Thus, using fractions of the components and relative humidity as variable parameters, the refractive index of a particle composed by several insoluble components (e.g., BC, BrC, mineral dust) suspended in such a host were determined by the MG equations based on the calculation of electric fields. Firstly, the complex dielectric constant of each aerosol component is computed from the corresponding complex refractive index. Secondly, the MG dielectric functions are used for calculating the dielectric constant of the aerosol mixture from component fractions and their dielectric constants. Lastly, the complex refractive index of aerosol mixture can be obtained from the complex dielectric constant of aerosol mixture.

Currently, the complete POLDER-3 data archive has been processed by the GRASP/Component approach to provide extensive retrievals of aerosol component concentrations and aerosol optical properties. Similar to previous products described in Chen et al. (2020), extensive GRASP/Component products generated as daily, monthly, seasonal, yearly, and climatological datasets in Level-2 and Level-3 are publicly accessible via GRASP-open (https://www.grasp-open.com/, last access: 20 July 2022). Although several regional applications of this GRASP/Component have been discussed in several studies (Li et al., 2020a, b; Zhang et al., 2021), the climatology of derived aerosol component distribution has not yet been fully analyzed in details, and therefore the full potential of this new satellite-derived aerosol composition was not fully clarified for the broad aerosol community. Therefore, the main focus of this study is on the analysis/verifications of aerosol composition (fractions) retrievals and considering the apparent climatological patterns of aerosol composition. Indeed, Zhang et al. (2021) validated the GRASP/Component optical properties against AERONET data and concluded that the generated total AOD values have minimal bias over both land (−0.02 for 550 nm) and ocean (0.01 for 550 nm), similar to the total AOD provided by GRASP/Models, while the detailed properties such as AE, AODF, and AODC have similarly good validation metrics as those provided by GRASP/HP. This suggests that the GRASP/Component products provide the overall most consistent total and detailed aerosol properties (e.g., spectral AODF, AODC, and
SSA) with respect to previous GRASP (Models, HP, and Optimized) products.

2.3 MERRA-2 data

The new version of the atmospheric reanalysis, MERRA-2, is publicly distributed by the NASA Global Modeling and Assimilation Office (GMAO) (Gelaro et al., 2017; Randels et al., 2017) and the detailed description can be found in (Buchard et al., 2017; Gelaro et al., 2017; Randels et al., 2017). The AOD products derived from the ground-based AERONET measurements and satellite observations (e.g., MODIS, MISR) are assimilated into MERRA-2 by the Goddard Earth Observing System (GEOS-5). The Goddard Chemistry, Aerosol, Radiation and Transport (GOCART) model is employed for the simulation of aerosol compositions such as BC, organic carbon, dust, etc. Up to now, numerous studies presented quantificationnal validations and assessments of different MERRA-2 products on a global or regional scale using measurements, such as AOD (Randels et al., 2017; Song et al., 2018; Sun et al., 2019), PM$_{2.5}$ (Buchard et al., 2016; Song et al., 2018), and surface BC concentration at several separate sites in China (Qin et al., 2019; Xu et al., 2020). However, MERRA-2 BC and dust column concentration products have not been compared to any observations globally. In this study, we employ the monthly MERRA-2 BC and dust column concentration products in a spatial resolution of $0.5^\circ \times 0.625^\circ$ as available from the NASA website (https://daac.gsfc.nasa.gov, last access: 20 July 2022).

3 Global POLDER-3 GRASP/Component aerosol products

The POLDER-3 GRASP/Component retrieval provides the following parameters characterizing the aerosol composition: volume fractions of BC, BrC, fine-/coarse-mode non-absorbing soluble and insoluble, coarse-mode absorbing, as well as fine-/coarse-mode AWC. From the view point of optical properties, the strongly absorbing and mostly scattering aerosol components are convenient to analyze separately.

3.1 GRASP/POLDER-3 absorbing component products

The following aerosol components determine the aerosol absorption: BC and BrC in fine-mode particles and CAI component mainly representing iron-oxide element contained in mineral dust.

3.1.1 Black Carbon (BC)

Figure 1 presents the distribution of seasonal BC column mass concentration in the climatological products derived from POLDER-3 observations using GRASP/Component, and Fig. 2 shows the corresponding standard deviations (SDs, calculated from data for all the months during the period 2005–2013, which is the same for other components in the following sections) divided by the mean of BC concentrations. These satellite-derived BC concentrations present noticeable temporal and spatial variations that are consistent with the distribution of biomass burning events. For example, over the African continent, the high BC concentrations can be observed over the Sahel region during the DJF season, extending to the region of southern Africa during the JJA and SON seasons as intense agricultural burnings occurring in the sub-Saharan region of Africa and progressing from north to south following the African monsoon cycle. In addition, large SD/mean values are obtained for BC concentrations over the region of southern Africa during the SON season. The elevated BC concentrations are also retrieved over South America during the entire biomass burning season (from August to November, see Fig. 1), which is in line with previous studies (Koren et al., 2007; Lei et al., 2021). At the same time, high SD/mean values for BC concentrations are obtained during the SON season over South America, which indicates the large interannual variability. Natural factors such as weather and cultural practices are known to make important contributions to the significant interannual variability of biomass burning in the Amazon basin rainforest (Koren et al., 2007). Figure 2 also indicates high SD/mean values for BC concentrations in Asia during the MAM season, which is associated with large interannual variability of fires in the biomass burning season (such as over the Indo-China Peninsula). Yearly means of BC column mass concentrations for the period from March 2005 to October 2013 is presented in Fig. 3. We should point out that many BC particles are generated from anthropogenic activities such as in China and India (shown in Fig. 4), however, our global climatology also indicates that the BC concentrations emitted from biomass burning in Africa and South America produce much higher BC concentrations than the anthropogenic emissions, e.g., in China and India (see Figs. 1 and 3). The climatology of the BC mass concentrations over the regions of East and South Asia are studied in detail by Li et al. (2020b), who demonstrated that GRASP/Component retrievals can characterize the temporal and spatial variations of light-absorbing carbon (BC and BrC) well in East and South Asia. A reasonable agreement between in situ ground-measured and satellite-derived BC mass concentrations in China has been demonstrated earlier by Li et al. (2020b). For instance, the BC products of GRASP/Component and the in situ measurements agree with the mean absolute difference of about 2.7 µm$^{-3}$, ranging from 0.09 to 7.8 µm$^{-3}$, and the relative difference of about 40 %, ranging from 2.6 % to 60 %, respectively (Li et al., 2020b). High BC proportions in Asia are found to be associated strongly with biomass burning events (such as agriculture waste open burning event during MAM over the Indo–China Peninsula or JJA in the North China Plain) and fossil fuel combustion (Li et al., 2020b). It is noted that high BC concentrations in India and China...
during the DJF season are strongly associated with anthropogenic activity (e.g., Li et al., 2019). The interpretation of some BC concentrations over the ocean, especially in high latitude near the polar region, should be taken with more caution because the sensitivity to the absorption and thus to the BC signal is limited when the aerosol loadings are very low and the retrievals are more uncertain, which is the same situation for large SD/mean values when BC concentrations are very low. In this respect, it can be noted that based on the sensitivity tests and uncertainty assessments in the study of Li et al. (2019), the uncertainty in BC fraction is within 50 % when AOD (440 nm) is larger than 0.4 and fraction is higher than 0.01. This BC uncertainty results mainly from the highly variable complex refractive index of BC, e.g., from 1.75+0.63i to 1.95+0.79i reported in the literature (Bond et al., 2013; Bond and Bergstrom, 2006). Similarly, the uncertainty in BC retrievals from ground-based AERONET measurements is about 50 % (Schuster et al., 2016).

### 3.1.2 Brown Carbon (BrC)

Unlike the strong and spectrally flat absorption of BC component, BrC shows a strong spectral dependence of absorption that decreases from shorter to longer wavelengths and a weaker absorption than BC (Chakrabarty et al., 2010; Kirchstetter et al., 2004; Wonaschütz et al., 2009). Figures 5 and 6 present the satellite-derived seasonal BrC column mass concentrations issued by GRASP/Component for the period 2005–2013 and the corresponding SDs divided by the mean of BrC concentrations, respectively. The seasonal variation and spatial distribution of BrC column mass concentrations show similarities to the BC pattern (in Fig. 1) as these light-absorbing carbonaceous particles are generally emitted from the same emission sources, such as biomass or fossil fuel combustion. Therefore, high BrC concentrations are also observed over Africa, South America, and the Indo–China Peninsula in Asia during the corresponding biomass burning period as discussed above for the BC concentrations. High BrC concentrations observed over the ocean near the continent is rather associated with the aerosol transport. For example, strong fires occurring from northern to southern Africa during the whole biomass burning period make large contributions to the burden of the BrC particles over the Atlantic Ocean near the African continent during the DJF, JJA, and SON seasons. Elevated BrC concentrations are also seen over the Pacific Ocean during the MAM season that can be transport of biomass burning emissions from the Indo–China Peninsula (Li et al., 2020b). The seasonal variabilities of BC and BrC concentrations in Asia observed in the global interannual climatology (Figs. 1 and 5), were already analyzed in the earlier study by Li et al. (2020b). Although statistically significant differences of smoke properties have been observed for fires in different fuel types such as forests, savannas, and grasslands (Junghenn Noyes et al., 2021), our component retrievals indicate that the elevated BC concentrations in this paper were attributed to fresh biomass burning aerosols, while those of the BrC concentrations were attributed to the aged biomass burning aerosols. Indeed, interactions with atmospheric gases (e.g., O3, NO2, and SO2) together with higher atmospheric water content over ocean can change the morphology and optical property of BC particles by oxidation and coating, e.g., (Adachi and Buseck, 2013; China et al., 2013; Decesari et al., 2002; J. Zhang et al., 2008). Therefore, the meteorological factors and the variability of atmospheric gases are also expected to contribute to the BrC seasonal and interannual variability. It can be seen from Fig. 6 that SD/mean values present no significant differences, which is similar to the small interannual variability of BrC concentrations in Fig. 7, except over southern Africa during the SON season (low concentration in 2012 and 2013). For proper data interpretation, it is worth noting that the uncertainty in BrC fraction is normally less than 50 % if the BrC fraction is above 0.1, even for very low AOD (smaller than 0.05), although the uncertainty is large in the case of small BrC fraction and low aerosol loading; the error estimations for GRASP/Component were conducted in Li et al. (2019).

### 3.1.3 Coarse-mode absorbing insoluble (CAI) aerosol component

Due to the absorption of hematite and goethite (free iron) contained in dust particles, mineral dust also indicates a spectral absorbing signature, i.e., stronger absorption at short wavelengths and weaker absorption at long wavelengths. In order to know the global distribution of free iron concentration for its important biogeochemical and radiative impacts, the CAI component was identified as an important species in the GRASP/Component approach (Li et al., 2019). At the same time, because of a similar tendency in spectral absorption for BrC and CAI, the size-resolved component configuration was selected, i.e., light-absorbing carbonaceous components (BC and BrC) are assumed to be absorbing components in fine-mode aerosol particles, and CAI mainly representing iron oxides is assumed to be an absorbing component in coarse-mode aerosol particles (Li et al., 2019). Figure 8 shows the satellite-derived seasonal CAI column mass concentration in the climatological products by GRASP/Component for the period 2005–2013. As discussed by Li et al. (2019), CAI particles mainly represent iron-oxide elements contained in mineral dust. From Fig. 8, we can see that elevated CAI concentrations are observed over the Sahara in western Africa, the Taklimakan Desert in northwest China, and over the Arabian Peninsula during the MAM and JJA seasons. These locations represent the global dust belt. Interesting to note is that very low CAI aerosol concentrations are obtained over the Bodélé Depression, which indicates high scattering dust particles (see the hot spot in Figs. 11 and 14 during the DJF season). This fact is consistent with the known dust composition nature of the Bodélé containing mainly dolomites (Formenti et al., 2011).
Figure 1. Spatial distribution of seasonal BC column mass concentration (mg m\(^{-2}\)) derived from POLDER-3 observations using the GRASP/Component approach. December–January–February (DJF); March–April–May (MAM); June–July–August (JJA); September–October–November (SON).

Figure 2. The standard deviation (SD)/mean for seasonal BC column mass concentration in Fig. 1 derived from POLDER-3 observations using the GRASP/Component approach. December–January–February (DJF); March–April–May (MAM); June–July–August (JJA); September–October–November (SON).
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Figure 3. Yearly mean of BC column mass concentration (mg m$^{-2}$) derived from POLDER-3 observations using the GRASP/Component approach for the period from March 2005 to October 2013.

Figure 4. Climatological BC column mass concentration (mg m$^{-2}$) derived from POLDER-3 observations using the GRASP/Component approach for the period from March 2005 to October 2013 in Asia.

The maximum mass ratio of about 5% of absorbing dust to scattering dust was found in our retrievals, which is consistent with the known mass fraction of iron oxides varying from 3% to 5% in the desert dust measurements (Guieu et al., 2002; Zhang et al., 2003). The characteristics presented in the climatological CAI distribution (Fig. 8) are also consistent with the results of previous studies (Formenti et al., 2008; Krueger et al., 2004; Lázaro et al., 2008) reporting large proportions of iron oxides in dust originating from the Sahel belt between 0 and 20° N and Saudi Arabia, but low iron-oxide content in the Chad Basin. The SD/mean of the seasonal CAI column mass concentrations (Fig. 9) generally show high values in the desert dust regions that are mostly due to the significant changes in 2012 and 2013 shown by yearly means of CAI (Fig. 10).

It should be recalled that the CAI particles are parameterized in the GRASP/Component approach as particles having spectral absorption of iron oxides, and this is the only absorbing species assumed in the coarse mode due to the limitations of measurement sensitivity. However, other light-absorbing species present in the coarse mode can produce similar signals and erroneously be interpreted as iron oxides. For instance, it can be noted that the CAI particles also appear in the regions and seasons associated with biomass burning, such as in South America during the SON season when strong biomass burning events occur. The CAI in this case is therefore an indicator that carbonaceous particles might be appearing in coarse mode by some mechanisms. Alternatively, some CAI particles appearing in South America may be originating in dust transport from western Africa. This is in line with the previous studies reporting the African dust transport over the Atlantic Ocean, Caribbean, Central America, and South America (during the DJF and MAM seasons) (Griffin et al., 2002; Kalashnikova and Kahn, 2008; Prospero and Lamb, 2003; Prospero and Mayol-Bracero, 2013). Anal-
Figure 5. Spatial distribution of seasonal BrC column mass concentration (mg m$^{-2}$) derived from POLDER-3 observations using the GRASP/Component approach. December–January–February (DJF); March–April–May (MAM); June–July–August (JJA); September–October–November (SON).

Figure 6. The standard deviation (SD)/mean for seasonal BrC column mass concentration in Fig. 5 derived from POLDER-3 observations using the GRASP/Component approach. December–January–February (DJF); March–April–May (MAM); June–July–August (JJA); September–October–November (SON).
3.2 GRASP/POLDER-3 scattering component products

As defined in Li et al. (2019), the weakly absorbing, mostly scattering aerosol components including the CNAI and FNAI components, as well as NAS and AWC in fine and coarse modes are discussed below.

3.2.1 Coarse-mode non-absorbing insoluble (CNAI)

The distribution of the seasonal CNAI aerosol component and the associated SD/mean are presented in Figs. 11 and 12, respectively. Yearly means of CNAI column mass concentration for the period from March 2005 to October 2013 is presented in Fig. 13. Generally, the CNAI component represents the non-absorbing part of mineral dust particles. Figures 11 and 13 reveal in detail the global dust belt with noticeable maximum concentrations during the MAM and JJA seasons. The high SD/mean values obtained over the main desert dust regions represent large interannual variability (Figs. 12 and 13). Specifically, the maximum CNAI concentration in northern Africa shows a shift from lower to higher latitudes during the MAM and JJA seasons, which is associated with the change of the intertropical convergence zone. The most active dust emission source, Bodélé Depression, is seen well in the CNAI retrievals (shown in Fig. 11) with the peaks during the DJF and MAM seasons, consistent with studies of Todd et al. (2007) and Washington et al. (2003). High coarse-mode non-absorbing dust concentrations are observed over the Arabian Peninsula and Taklamakan in China during the MAM and JJA seasons, while a significant decrease appears during the SON and DJF seasons. In addition, a key dust emission source in southwest Asia, named the Sistan region (Goudie, 2014; Léon and Legrand, 2003; Middleton, 1986), can be seen in Fig. 11 during the JJA season.

The well-known locations of source regions of mineral dust across the globe are reported by numerous studies: the Sahara region in northern Africa (e.g., Choobari et al., 2014; Prospero et al., 2002); the Middle East and Arabian Peninsula (Ginoux et al., 2012; Prospero et al., 2002); the Sistan Basin (representing the region of Iran, Pakistan, Afghanistan); the desert regions in southwestern Asia.
Figure 8. Spatial distribution of seasonal coarse-mode absorbing insoluble (CAI) column mass concentration \((\text{mg} \, \text{m}^{-2})\) derived from POLDER-3 observations using the GRASP/Component approach. December–January–February (DJF); March–April–May (MAM); June–July–August (JJA); September–October–November (SON).

Figure 9. The standard deviation (SD)/mean for seasonal CAI column mass concentration in Fig. 8 derived from POLDER-3 observations using the GRASP/Component approach. December–January–February (DJF); March–April–May (MAM); June–July–August (JJA); September–October–November (SON).
calculated yearly mean of CAI column mass concentration (mg m$^{-2}$) derived from POLDER-3 observations using the GRASP/Component approach for the period from March 2005 to October 2013. The Taklamakan in China and Gobi (northern China–southern Mongolia) deserts (Ginoux et al., 2012); as well as several desert regions in central Asia (Elguindi et al., 2016) and in North America (Ginoux et al., 2012). All of them represent the Northern Hemisphere as the location with the most important natural dust emission sources. To date, the studies of these dust sources were conducted based on satellite remote sensing, mostly providing the total AOD, AODC or AOD related, under extra assumptions, to desert Dust Optical Depth (DOD) (Logothetis et al., 2021). Those studies were conducted based on derived dust indices, AOD, aerosol profiles, and on thermal infrared signal, i.e., from products of the Total Ozone Mapping Spectrometer (TOMS; Ginoux et al., 2012; Prospero et al., 2002), MODIS (Gui et al., 2022; Remer et al., 2008; Schepanski et al., 2012; Song et al., 2021; Voss and Evan, 2020; Yu et al., 2019), Infrared Atmospheric Sounding Interferometer (IASI; Clarisse et al., 2019; Yu et al., 2019), MISR (Gui et al., 2021a; Kahn and Gaitley, 2015; Yu et al., 2019), and Cloud–Aerosol Lidar with Orthogonal Polarization (CALIOP; Gui et al., 2021b, 2022; Shikwambana and Sivakumar, 2018; Song et al., 2021). In this regard, the GRASP/Component product from POLDER-3 privies certainly more direct quantitative evaluations of the mineral dust composition and its iron-oxide content, enabling studies of the trends and variability. For example, as presented in Fig. 13, over the Arabian Peninsula, the lower CNAI concentrations are obtained for 2005–2007 and higher values of CNAI concentrations are obtained for the 2008–2013 period, which agrees well with the results of Notaro et al. (2015) who indicated inactive and active dust periods (1998–2005 and 2007–2013, respectively) over the Arabian Peninsula related to the sustained drought near the Fertile Crescent region. The MODIS DOD products indicated that the dust loadings have an increase over the Sahara region during the period 2003–2018 (Voss and Evan, 2020), which can now be seen quantitatively in Fig. 13. We should also point out that some CNAI particles observed near central southern Africa might be attributed to the CNAI organic particles that probably exist when and where there are strong carbonaceous aerosol emissions during the biomass burning events (in Fig. 13).

3.2.2 Fine-mode non-absorbing insoluble (FNAI)

Figure 14 shows the pattern of the seasonal FNAI column mass concentration in the derived climatology, as well as the SD divided by the mean of the FNAI concentration in Fig. 15. As mentioned previously and discussed by Li et al. (2019), both scattering organic carbon and non-absorbing dust particles in fine-mode aerosol are labeled as FNAI in current GRASP/Component. Thus, significant similarities and differences are obviously seen between the spatiotemporal distributions of FNAI (Fig. 14) and CNAI (Fig. 8). Specifi-
Figure 11. Spatial distribution of seasonal coarse-mode non-absorbing insoluble (CNAI) component column mass concentration (mg m$^{-2}$) derived from POLDER-3 observations using the GRASP/Component approach. December–January–February (DJF); March–April–May (MAM); June–July–August (JJA); September–October–November (SON).

Figure 12. The standard deviation (SD)/mean for seasonal CNAI column mass concentration in Fig. 11 derived from POLDER-3 observations using the GRASP/Component approach. December–January–February (DJF); March–April–May (MAM); June–July–August (JJA); September–October–November (SON).
Figure 13. Yearly mean of CNAI column mass concentration (mg m$^{-2}$) derived from POLDER-3 observations using the GRASP/Component approach for the period from March 2005 to October 2013.

cally, similar to CNAI and CAI, the spatiotemporal characteristics of the global dust belt appears in FNAI, i.e., the concentrations appear over northern Africa, Bodélé Depression, Arabian Peninsula, and the Taklimakan Desert. However, the differences are manifested in well-known locations for the biomass burning regions and seasons, i.e., in southern Africa and South America during the JJA and SON seasons, in the Indo–China Peninsula during the MAM season, as well as fossil fuel burning associated with strong anthropogenic activities in China and India. The retrieved FNAI particles, mainly representing the fine-mode scattering organic carbon, present good spatiotemporal consistency with the results of previous studies on the variability of anthropogenic and biomass burning emissions (Duncan et al., 2003; X. Y. Zhang et al., 2008, 2012). Large SD/mean values are obtained over southern Africa and South America, which can be attributed to the interannual variations of biomass burning events. For example, the yearly means of the FNAI column mass concentration shown in Fig. 16 presents extremely high FNAI concentration over South America in 2005, 2007, and 2010, which could be caused by the areas and frequency of El Niño-induced droughts and fires in this region. The uncertainty of the FNAI retrievals depends on the aerosol loading, and is generally less than 100 %.

3.2.3 Non-absorbing soluble (NAS) and aerosol water content (AWC)

As mentioned previously and discussed in Li et al. (2019), the components contained in the host can represent all ammonium nitrate, sulfate, sea salt and AWC in fine- and coarse-mode aerosol fraction. However, NAS can also be aged hydroscopic components, such as water-soluble organic carbon generated from biomass burning or anthropogenic activities, and water-soluble mineral ions provided by dust particles. The NAS and AWC components can thus be quite complex for interpretation, however, overall the obtained climatology presents a reasonable logic. For example, the fine-mode non-absorbing soluble (FNAS) component (Fig. 17) is high in China and India, especially during the DJF and SON seasons, which is in line with strong anthropogenic emissions (e.g., industrial and heating activities) in megacities where the population is dense. The non-absorbing water-soluble organic carbon, that can be emitted during biomass burning events, can also be detected and interpreted as FNAS, e.g., as it occurs in southern Africa. The GRASP/Component retrievals also indicate many NAS particles present in the Mediterranean region, which is in line with the previously demonstrated aerosol mixture constituted of anthropogenic aerosols, sea-salt particles, and mineral dust coated with biogenic sulfate (Ganor et al., 2000; Lelieveld et al., 2002; Levin, 2005; Levin et al., 1996). The spatiotempo-
Figure 14. Spatial distribution of seasonal fine-mode non-absorbing insoluble (FNAI) component column mass concentration (mg m\(^{-2}\)) derived from POLDER-3 observations using the GRASP/Component approach. December–January–February (DJF); March–April–May (MAM); June–July–August (JJA); September–October–November (SON).

Figure 15. The standard deviation (SD)/mean for seasonal FNAI column mass concentration in Fig. 14 derived from POLDER-3 observations using the GRASP/Component approach. December–January–February (DJF); March–April–May (MAM); June–July–August (JJA); September–October–November (SON).
Figure 16. Yearly mean of FNAI column mass concentration (mg m$^{-2}$) derived from POLDER-3 observations using the GRASP/Component approach for the period from March 2005 to October 2013.

Rat distributions of AWC appear to be similar to the NAS component, which is quite logical because these two components are naturally related and also form the particle host in the employed MG effective medium approximation mixing rule in the algorithm. The fine-mode AWC (Fig. 18) is also naturally associated with hygroscopic anthropogenic aerosol particles and dominates in regions such as Asia. The aerosol water presence over ocean during the dust transport and near southern Africa during the biomass burning is also rather logical because of higher atmospheric relative humidity. Moreover, a significant increase of mineral dust hygroscopicity was attributed in several studies to the aging processes and the dust mixing with soluble hygroscopic material (Sullivan et al., 2009; Tang et al., 2016). The impact of such mixtures on remote-sensing observations was also observed (Derimian et al., 2017; Falkovich et al., 2004). Besides some regions with high CNAI (dust) concentration over ocean (near western Africa and Arabian Peninsula), the values of NAS and AWC in coarse mode are rather low everywhere. We note that a relatively low fraction, but with very large volume concentration, could produce relatively important concentration. In addition, in situations when the real part of aerosol refractive index is very low, the algorithm may try to introduce a water fraction and thus produce an artifact. This is the case when a water fraction is retrieved over the Bodélé Depression during the DJF season. Indeed, an important proportion of fossil diatom in the sediments of the Bodélé Depression (Formenti et al., 2008) can produce dust with an unusually low real part of refractive index. To reproduce the optical signal of this dust, a small fraction of water can be required (element with the lowest real refractive index). Since the dust concentrations are extremely high in this source, even a small fraction presents relatively high water concentrations, which is an artifact. This example illustrates that interpretations of the retrieved component fractions are not always straightforward and depend on limits in the measurement sensitivity. Nevertheless, the measurements reflecting physical properties and a closer analysis can usually explain mismatches with expectations.

4 Comparisons of BC and dust concentration derived from GRASP/Component with MERRA-2

Although applications of GRASP/Component to different instruments (sun-photometer, POLDER-3, DPC/GF-5) have presented good consistency with the available AERONET aerosol optical products such as AOD, AAOD, AE, and also agree with the physical expectations about aerosol composition in different locations, additional reference data are still desirable for the validation. Because the in situ measurement data are extremely scarce and difficult for global validation, we employ an intercomparison of BC and dust column concentration between the GRASP/Component climatological dataset and the corresponding MERRA-2 prod-
Figure 17. Spatial distribution of seasonal fine-mode non-absorbing soluble (FNAS) component column mass concentration (mg m\(^{-2}\)) derived from POLDER-3 observations using the GRASP/Component approach. December–January–February (DJF); March–April–May (MAM); June–July–August (JJA); September–October–November (SON).

Figure 18. Spatial distribution of seasonal fine-mode aerosol water content (FAWC) column mass concentration (mg m\(^{-2}\)) derived from POLDER-3 observations using the GRASP/Component approach. December–January–February (DJF); March–April–May (MAM); June–July–August (JJA); September–October–November (SON).
ucts. Indeed, this is not a pure validation experiment, but can still provide a better indicator of a global consistency. The CNAI component here will refer to the dust component in MERRA-2. The GRASP/Component dust as well as BC products will be intercompared. To this end, the GRASP/Component products are re-gridded into the same resolution of 0.5° × 0.625° as that of MERRA-2. Figure 19 presents the differences of monthly BC column concentration between GRASP/Component and MERRA-2 for the period March 2005–October 2013. One can note that both GRASP/Component BC and MERRA-2 BC represent quite similar spatiotemporal patterns, and generally show reasonable consistency between them over ocean, Europe, United States, and mostly over South America and Africa. There are however some noticeable differences during the strong biomass burning season and strong anthropogenic emissions in China and India. In further discussions, we analyze the possible source of these discrepancies.

Firstly, MERRA-2 includes the estimations of several externally mixed aerosol components (BC, OC, dust, etc.) as defined in GOCART (Chin et al., 2002). The emissions for carbonaceous aerosol species in GOCART include fossil fuel combustion, biomass burning, biofuel consumption, and biogenic sources. Both natural and anthropogenic sources make contributions to BC emissions in MERRA-2. Various inventories during the different time periods were considered to produce biomass burning emissions of carbonaceous aerosols in MERRA-2. Specifically, the Quick Fire Emissions Dataset (QFED) version 2.4-r6 (Darmenov and da Silva, 2015) has been employed for daily biomass burning BC emissions since 2010. The monthly biomass burning emissions in the Global Fire Emission Dataset (GFED) version 3.1 (Randerson et al., 2006; Van Der Werf et al., 2006) were used for the period 1997–2009. Biome-dependent factors are used to make corrections between GFED v3.1 and QFED v2.4-r6 during the period 2003–2011 when they were available. On the other hand, it is noted that more and more strict controls on air pollution in China have taken place with the rapid economic growth (Zhang et al., 2019). The inventory of Peking university (PKU) indicated anthropogenic aerosol emissions present a decrease in the key regions of China since 2006, with a significant trend of −1.4% per year for anthropogenic BC emissions during the period 2006–2014 (Wang et al., 2021). However, a continuous increase of anthropogenic aerosol emissions in China from 2006 to 2014 is shown in the Community Emissions Data System (CEDS) inventory, thus the global climate models cannot capture the observed downward trend well (Wang et al., 2021). The comparisons of biomass burning emission datasets, including GFED version 3.1 (GFED3.1), GFED version 4 with small fires (GFED4s), Fire Inventory from NCAR version 1.5 (FINN1.5), Global Fire Assimilation System version 1.2 (GFAS1.2), Fire Energetics and Emissions Research version 1.0 (FEER1.0), and Quick Fire Emissions Dataset version 2.4 (QFED2.4) for 2008 globally have indicated that a factor of 3.4 can be obtained on an annual average for the differences of BC emissions (Pan et al., 2020). At the same time, the knowledge on the complexity and variability of different fire features are still required to be improved (Hyer et al., 2011). Continuous measurements are required to contribute toward advancing our understanding of the fire-generated BC globally, at least in the major biomass burning regions. The emission factors for different ecosystem types and burning stages need to be estimated accurately for the improvement of emissions globally (Pan et al., 2020).

Secondly, a higher value of the BC refractive index (1.95–0.79i) employed in GRASP/Component for the BC retrievals (Li et al., 2019) than the BC refractive index (1.75–0.45i) used in the GOCART model (Chin et al., 2002), which affects the GRASP BC mass retrievals, also makes contributions to the BC differences. Finally, we should also note that a new light-absorbing carbonaceous component, named BrC, is included in the GRASP/Component products so that the spectral aerosol absorption signature in the fine-mode fraction can be interpreted as BrC contribution in the satellite inversion. At present, MERRA-2 does not provide the BrC product. Therefore, the noticeable BrC concentrations for the strong biomass burning season (in Fig. 5), over ocean near southern Africa, can indeed explain the BC underestimation by GRASP/Component relative to MERRA-2. The comparisons of GOCART BrC estimations (Schill et al., 2020) and GRASP BrC retrievals are expected to be done in a future study.

The monthly dust column concentration differences between GRASP/Component and MERRA-2 for the period March 2005–October 2013 are presented in Fig. 20. We can see that GRASP/Component dust concentration is less than one of MERRA-2 over northern Africa, Arabian Peninsula, Middle Asia, and northwest China. We would like to recall that the wind speed-dependent emissions with the potential dust source locations described in Ginoux et al. (2001) are used for dust missions in the MERRA-2/GOCART model, and that the depositions (such as dry deposition, wet removal globally) of aerosol are not considered in the current GOCART estimation for MERRA-2 (Randles et al., 2017). These factors can cause the overestimation of dust in MERRA-2. At the same time, the GRASP/Component dust concentration used for the comparison represent the CNAI component as it is defined in MERRA-2. These differences can merit a closer future analysis as a function of dust source region, composition and transport distance. We note that the GRASP CNAI retrieval shows slightly higher dust in southern Africa than MERRA-2, which might be related to the possible existing coarse-mode organic particles during the biomass burning seasons with strong carbonaceous aerosol emissions.
Figure 19. Differences of monthly BC column concentration between GRASP/Component and MERRA-2 in the resolution of 0.5° × 0.625° for the period March 2005–October 2013.

5 Data availability


6 Conclusions

The current paper presents a climatological analysis of the aerosol product recently generated from POLDER-3 using the newly developed GRASP/Component approach. This approach employs an essentially different methodology, compared to previous algorithms, and attempts to derive not only optical properties of aerosol, but also some information about aerosol composition. The GRASP/Component algorithm provides global satellite observations based on concentrations of absorbing components such as BC, BrC, and CAI representing iron oxides contained in dust, and of scattering components such as fine-mode organic carbon/scattering dust, and coarse-mode scattering dust, which is scarce but imperative information required for tuning the chemical transport models. This GRASP-generated, satellite-derived aerosol component product is expected to be valuable for qualitative and quantitative understanding of aerosol component distribution on global and regional scales. More importantly, the data assimilation of this extensive satellite-based aerosol component dataset can importantly contribute to improving the emissions of global aerosol estimation, and the further improvement of accuracy of the estimated aerosol radiative forcing in general and per aerosol component in particular. For instance, the AOD and AAOD products de-
derived from POLDER-3 observations have already been used to constrain GEOS–Chem inverse modeling for the improvement of global BC, organic carbon and desert dust aerosol emissions (Chen et al., 2018, 2019). Using the additional satellite-based aerosol component products presented in this study, the further improvement of global aerosol emissions estimation is expected. The presented efforts are also in line with the studies suggesting that the employment of satellite-constrained anthropogenic and natural aerosol emissions by the climate models is required to improve the accuracy of aerosol radiative forcing estimations (e.g., Bellouin et al., 2020a; Quaas et al., 2022). That is, the linear trends in column concentration of the main aerosol components, such as BC, BrC, CAI and CNAI as shown in Fig. S1 in the Supplement, can provide better satellite-measured constraints on the properties of anthropogenic (BC and BrC) and natural (CAI and CNAI) aerosols on a global scale, and will contribute to improving the accuracy of anthropogenic aerosol radiative forcing estimations. This paper shows and discusses the climatological patterns derived from POLDER-3 using the GRASP/Component approach. It is shown that this global satellite-derived climatology of aerosol chemical composition clearly reveals the distribution of global dust belt, biomass burning events, and anthropogenic activity emissions that are in good agreement with known logical expectations and previous studies. The intercomparisons of BC and dust column concentrations provided by GRASP/Component and MERRA-2 for the period March 2005–October 2013 show good agreement in global spatiotemporal patterns, although some explainable differences appear for strong biomass burning events and in regions strongly affected by mineral dust. For example, the BC emissions can significantly affect the MERRA-2 BC concentrations during the strong biomass burning periods.

In addition, all GRASP/Component-generated optical properties agree very closely with the best previously...
obtained POLDER-3 aerosol products, not only over AERONET stations, but also globally, over both land and ocean. Specifically, the GRASP/Component approach is demonstrated to provide the aerosol optical characteristics of comparison to or sometimes even better accuracy (as validated by AERONET data) than all other earlier GRASP/Optimized, HP and Models products. For example, the validation study by Zhang et al. (2021) using global AEROENT data showed that total spectral AOD quite closely agrees with GRASP/Models retrievals that has the highest correlation and lowest bias (Chen et al., 2020). At the same time, most of the GRASP/Component detailed characteristics (spectral AODF, AODC, AAOD and SSA) provided by GRASP/Component also agree well with the aerosol optical products of previous GRASP approaches (Optimized, HP, and Models) in Chen et al. (2020).

Although a rigorous global validation of aerosol components remains challenging at the current stage, its estimation is the next frontier for aerosol inversion from satellite observations, because it is required for continuous evaluation of the aerosol life cycle by chemical transport models. Also, it should be noted that in situ aerosol component data are scarce, unlike satellite AOD products on global scale. The satellite aerosol component product presented here is comparable in coverage with the conventional AOD product, while retrieval requires more complex aerosol parameterization and advanced remote-sensing observations, such as the MAP observations. At the same time, many spaceborne instruments with the multi-angle polarization measurement capabilities are planned to be deployed and available in the next few years (Dubovik et al., 2019), such as Multi-Channel Multi-Polarization Imaging (3MI), Directional Polarimetric Camera (DPC-02), etc. Finally, the presented paper illustrates that with the synergy of different remote-sensing instruments and additional algorithm refinements, the extensive retrievals of aerosol, microphysical, and optical properties together with aerosol composition information, as derived from the GRASP/Component algorithm, can be generated extensively in the near future.

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