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Non-spherical microparticle shape in Antarctica during the last glacial period affects dust volume-related metrics

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Abstract. Knowledge of microparticle geometry is essential for accurate calculation of ice core volume-related dust metrics (mass, flux, and particle size distributions) and subsequent paleoclimate interpretations, yet particle shape data remain sparse in Antarctica. Here we present 41 discrete particle shape measurements, volume calculations, and calibrated continuous particle time series spanning 50-16 ka from the South Pole Ice Core (SPC14) to assess particle shape characteristics and variability. We used FlowCAM, a dynamic particle imaging instrument, to measure aspect ratios (width divided by length) of microparticles. We then compared those results to Coulter counter measurements on the same set of samples as well as high-resolution laserbased (Abakus) data collected from SPC14 during continuous flow analysis. The 41 discrete samples were collected during three periods of millennial-scale climate variability: Heinrich Stadial 1 (18–16 ka, n = 6; ~ 250 years per sample), the Last Glacial Maximum (LGM) (27–18 ka, n = 19; \sim 460 years per sample), and during both Heinrich Stadial 4 $(42-36 \text{ ka}, n = 8; \sim 620 \text{ years per sample})$ and Heinrich Stadial 5 (50–46 ka, n = 8; ~ 440 years per sample). Using FlowCAM measurements, we calculated different particle size distributions (PSDs) for spherical and ellipsoidal volume estimates. Our calculated volumes were then compared to published Abakus calibration techniques. We found that Abakus-derived PSDs calculated assuming ellipsoidal, rather than spherical, particle shapes provide a more accurate representation of PSDs measured by Coulter counter, reducing Abakus to Coulter counter flux and mass ratios from 1.82 (spherical assumption) to 0.79 and 1.20 (ellipsoidal assumptions; 1 being a perfect match). Coarser particles ($> 5.0 \,\mu m$ diameter) show greater variation in measured aspect ratios than finer particles ($< 5.0 \,\mu$ m). While fine particle volumes can be accurately estimated using the spherical assumption, applying the same assumption to coarse particles has a large effect on inferred particle volumes. Temporally, coarse and fine particle aspect ratios do not significantly change within or among the three time periods (p value > 0.05), suggesting that long-range transport of dust is likely dominated by clay minerals and other elongated minerals.

1 Introduction

Ice core microparticle (i.e., insoluble dust) data provide a critical perspective on past climate variability because a

range of dust metrics (e.g., number and mass concentrations, flux, volume, size distribution, and geochemical composition) can be measured and used to reconstruct atmospheric circulation (i.e., Delmonte et al., 2002; Koffman et al., 2014; Aarons et al., 2017; Wegner et al., 2015; Lambert et al., 2012; Petit et al., 1999), radiation balance (i.e., Lambert et al., 2013; Durant et al., 2009; Baggenstos et al., 2019), and chemical delivery to ocean and terrestrial ecosystems (e.g., Conway et al., 2015; Edwards et al., 2006; Gaspari et al., 2006; Spolaor et al., 2013; Vallelonga et al., 2013; Wolff et al., 2006). There are two analytical techniques commonly used to measure dust particle size and concentration in ice cores: Coulter counter (CC) and continuous laser particle sensor (Abakus). Each of these methods involve assumptions regarding particle shape. The CC measures particle concentration through electrical resistance in discrete samples, whereby a solution is fed through an aperture which disrupts a constant electrical field between two electrodes. The impedance of the electrical signal is directly proportional to the particle volume, and therefore CC data are generally assumed to be the most accurate estimate of particle volume. The Abakus measures particles via disruption of transmitted light; as particles flow through the sensor, the interruption of the light source produces scattering and shadowing, resulting in a negative peak in transmitted light, which is used to measure the diameter of the particle (Ruth et al., 2002). A main advantage of the Abakus is that it can be used in a continuous flow analysis (CFA) system, providing theoretical millimeter-scale resolution and allowing for in-line comparison with other CFA ice core data (e.g., soluble ions, black carbon, stable isotopes, electrical conductivity). However, it only provides particle geometric size measurements in one dimension (i.e., particle length or width) and does not provide depth or height information for individual particles (Ruth et al., 2002; Simonsen et al., 2018). Abakus data have traditionally been interpreted using the assumption that the measured length of each particle is equal to the diameter of that particle and that all microparticles are spherical. While both approaches have advantages, neither provides direct information about particle shape.

Recent work has tested assumptions about particle sphericity by using a third technique, the single-particle extinction and scattering (SPES) method, and found that the spherical approximation leads to mismatches not only in volume but also in calculated particle size distribution (PSD) between Abakus and CC data (Simonsen et al., 2018; Villa et al., 2016; Potenza et al., 2016, 2015). SPES uses particle light scattering and absorption measurements to identify extinction cross-sections. Particles are directed through the focal point of a light beam while sensors in the far field collect the total power removed from light scattering and absorption of the particle. Because the sum of scattering and absorption by each particle is proportional to the extinction crosssection, particle shape information can be identified for particles relatively quickly and efficiently (Potenza et al., 2015). Using a combination of SPES, CC, and Abakus methods, Simonsen et al. (2018) provided the first calibration scheme to correct for offsets in the PSD of Abakus-derived data based on particle shape for the Holocene and last glacial period in Greenland.

We expand on previous work by developing and presenting here new dust data from the South Pole Ice Core (SPC14) that spans 50-10 ka and includes Abakus, CC, and particle shape measurements. We identified three intervals of interest (Heinrich Stadial 1, HS1, 18-16 ka; the Last Glacial Maximum, LGM, 27-18 ka; and Heinrich Stadial 4 and Heinrich Stadial 5, HS4 and HS5, 42-36 and 50-46 ka, respectively) based on periods of millennial-scale climate changes in stable isotope, microparticle concentration, and CO₂ data (Fig. 1). SPC14 has a relatively high accumulation rate $(7.4 \text{ cm yr}^{-1} \text{ during the Holocene and about } 3.7 \text{ cm yr}^{-1} \text{ dur-}$ ing the LGM) and thus a higher temporal resolution compared with many other sites in East Antarctica (Winski et al., 2019, 2021; Casey et al., 2014; Lazzara et al., 2012; Kahle et al., 2020). To assess the size- and time-dependent variability in particle geometry in SPC14, we utilize dynamic particle imaging (DPI) to measure particle shape properties (length, width, and aspect ratio; width / length). DPI via the flow cytometer and microscope (FlowCAM) is a technique that allows for rapid counting, imaging, and measurement of particles (Sieracki et al., 1998) and is advantageous for ice core microparticle analyses because of its direct imaging capabilities and fast data output. DPI techniques have been extensively used in the biology community to analyze plankton species in a variety of environments (Álvarez et al., 2011). We use DPI at selected depth intervals in SPC14 to obtain aspect ratio measurements and then apply these data to continuous Abakus and discrete CC measurements in order to assess interpretations of the SPC14 microparticle records. Depths were selected based on millennial-scale Abakus dust concentration and size variability recorded in the preliminary SPC14 ice core. In addition, we combine DPI and Abakus data to calculate particle volumes and compare them against published particle shape calibration techniques from Simonsen et al. (2018). Our analyses provide the first millennially resolved time series of particle shapes and SPC14 microparticle data spanning the interval 50–10 ka, providing valuable information regarding particle shape variability during the last glacial period and the start of the last termination in Antarctica.

2 Methods

2.1 Ice core recovery

SPC14 was drilled by the US Ice Drilling Program with the Intermediate Depth Drill (IDD) using a 98 mm drill head (Winski et al., 2019; Johnson, 2014; Epifanio et al., 2020). Drilling fluid (Estisol-140) was introduced after reaching a depth of 160 m to reduce frictional resistance and to keep



Figure 1. (a) Ages of discrete particle shape analyses in this study compared to previously published research on particle shape variability during the last glacial period (LGP): Potenza et al. (2016; Dome C, Antarctica; blue); Simonsen et al. (2018; Renland Ice Cap, Greenland; red); SPC14 CC (present study; green); and SPC14 DPI (present study; purple). Grey SPC14 CC and DPI markers represent samples that were run (see Methods). (b) SPC14 dust particle concentration (present study; light brown), (c) δ^{18} O from Steig et al. (2021; grey), and WAIS Divide CO₂ (Bauska et al., 2021; Marcott et al., 2014; salmon) are shown for context. Grey bars highlight time periods of interest in our study.

the borehole open. Core sections from the brittle ice zone (BIZ; 585–1077 m) were stored at the site for 1 year after drilling to depressurize before shipment to the US. Core depth, weight, and density measurements were completed at the US National Science Foundation Ice Core Facility (NSF-ICF) in Denver, Colorado. Sections of SPC14 were cut into $2.4 \text{ cm} \times 2.4 \text{ cm} \times 100 \text{ cm}$ "sticks" at NSF-ICF and shipped frozen to Dartmouth College.

2.2 Abakus measurements

High-resolution Abakus measurements were made at Dartmouth College using a continuous flow analysis (CFA) melter system (Osterberg et al., 2006; Breton et al., 2012). Core sticks were melted on a 99.9995% pure chemicalvapor-deposited silicon carbide (CVD-SIC) melt head. The Dartmouth College CFA system has an effective resolution of 3 mm with signal dispersion lead of ~ 1 cm (Breton et al., 2012; Winski et al., 2019; Koffman et al., 2014; Osterberg et al., 2006). Microparticles were measured using a Klotz Abakus laser particle detector (Abakus). Peristaltic pumps drew meltwater from the ice core through tubing directly into the Abakus sensor. Abakus particle concentrations (particles μL^{-1}) were calculated by dividing the Abakus output (particle count per unit time; usually 6 s) by the melt flow rates, which were measured using a flowmeter (Sensirion SLI-2000). A total of 24 Abakus size bins ranged continuously in 0.1 µm increments from endpoints (lower) 1.0– 2.5 μ m, followed by bins 2.7, 2.9, 3.2, 3.6, 4.0, 4.5, 5.1, 5.7, and 6.4 μ m. Particle mass concentrations are estimated using an assumed density of 2.6 g cm⁻³ (i.e., following Koffman et al., 2014). We calculate mass and flux measurements using spherical and ellipsoidal volume with values obtained from DPI measurements (see Sect. 2.4 and 2.5). Age data were interpolated based on the SP19 timescale (Epifanio et al., 2020; Winski et al., 2019). The Abakus was calibrated using CC techniques and then successfully tested against latex spheres of 1, 2, 5, and 10 μ m diameter to assess accuracy (Koffman et al., 2014). Therefore, our size range of 1.1–6.4 μ m has been successfully used and is considered accurate.

To check if the coarse particle concentration measured by the Abakus is composed of large particles rather than multiple small coagulated particles, we tested for coincidence following methods of Saey (1998) and used in Simonsen et al. (2018). Saey (1998) showed that during measurement of ice with high particle concentration (> $240\,000\,\text{particles mL}^{-1}$) two particles passing through the detector at the same time can be erroneously identified as one large particle (hereafter referred to as "particle coincidence"). Following methods of Simonsen et al. (2018), we compared the ratio of two different metrics of coarse particles (5.1–6.4 and 3.2–6.4 μ m) to fine particles (< 5.1 and $< 3.2 \,\mu$ m, respectively) against particle concentration during our periods of highest Abakus particle concentration using a Spearman correlation test. If there was a high correlation of coarse particles with particle concentration, we would infer that particle coincidence occurred during CFA data collection. While we are aware of particle aggregation in deep sections of ice cores, we also access the impact of aggregation on PSD values. Baccolo et al. (2018) identified particle aggregates in the deep sections of the Talos Dome core PSDs, which were characterized by low fine particle concentration and a coarse ($\sim 4.4 \,\mu\text{m}$) mode. We look for similar features in the deep South Pole Ice Core.

Abakus data cleaning

Previous studies have highlighted the contaminating effect of the drilling fluid Estisol-140 on ice core microparticle concentrations (Warming et al., 2013). Estisol-140 peaks are characterized in the microparticle concentration data by exponentially increasing particle concentration followed by a sharp decrease of several orders of magnitude back to the background concentration (Fig. S1 in the Supplement). Because of the repeatable nature of each Estisol-140 contamination peak, we were able to remove spurious data from our high-resolution Abakus microparticle dataset using a Python-based cleaning code (Anderson, 2020). The code uses changes in coarse particle percentage (CPP) that overlap with changes in the rolling (three-point) median. If the metrics overlap, the data points in those time periods are flagged for removal. The cleaning code also utilizes the highresolution Abakus data as well as multiple metrics collected in-line during CFA melting (i.e., electrical conductivity measurements and flow rates) and targets sections of the core where issues occurred during CFA melting and data collection as well as where erroneous peaks occurred from contamination of Estisol-140. In total, this approach indicates that, at most, 3% of the SPC14 record contains microparticle data affected by Estisol-140 (Fig. S1). Similar quality control cannot be performed on the CC data because there are only 54 CC samples (see next section) compared with several hundred thousand Abakus measurements.

2.3 Coulter counter measurements

During melting on the CFA system, discrete samples were collected every 2 m (866 samples total; \sim 125 years per sample age span during the LGM) for glaciochemical analysis. Of these, we selected 106 to combine into 54 low-resolution (2-4 m; see the Supplement) CC samples for comparison with the Abakus data (Fig. 1; Supplement). CC samples were analyzed at Colby College using a Beckman Coulter Multisizer 3 in a HEPA-filtered clean lab. To prepare a sample for analysis, we poured 10 mL from each of the 54 samples into cuvettes that were cleaned with a 5 % Citranox solution and multiple rinses with Milli-QTM ultrapure water. The samples were then mixed with Isoton IITM diluent (electrolyte solution) and inverted several times prior to analysis. Samples were measured using a 50 µm aperture tube, yielding measurements of particle concentration across 300 size bins spanning 1.1–30 µm. Three to five blank measurements were made at the start of each sampling day. Following clean blank runs, three to five measurements were made for each of the 54 samples. Average blank measurements were about 160 times lower than raw CC data and were significantly different from sample measurements according to a Student's t test. We performed blank corrections by subtracting the averaged median blank value from the median of each CC sample, since blank values are within 2σ of each other. Because the CC produces more size bins than the Abakus (300 vs. 30 bins), we averaged CC particle size bin concentrations to match Abakus bin sizes for direct comparisons.

2.4 Dynamic particle imaging

The FlowCAM (Yokogawa Fluid Imaging Technologies, Inc.) instrument at the University of Maine uses DPI to measure particle length, width, and aspect ratios. DPI via Flow-CAM was originally developed to photographically measure phytoplankton or other particles $20-200 \,\mu\text{m}$ in length in concentrations ranging from $10^1 - 10^5 \,\text{L}^{-1}$ (Sieracki et al., 1998). Modern DPI techniques using a $20 \times \text{ zoom}$ can accurately image particles $\geq 2.0 \,\mu\text{m}$ in length, while recording particle shape (length and width) information throughout each sampling run. Particle length, width, and aspect ratio measurements were calculated using Feret measurements, defined as the perpendicular distance between parallel tangents on op-

posite sides of the particles (Fluid Imaging Technologies, 2011). These measurements are taken 36 times in 5° angle increments from -90° to $+90^{\circ}$. The maximum distance recorded is the particle length and the minimum is recorded as the particle width. Aspect ratios are then calculated by the width (b axis) / length (a axis). Values range between 0 and 1, where 0 represents elongated (oblate and/or prolate) particles and 1 represents a spherical particle. While 2D imaging does provide a length and width measurement, this technique is limited because we cannot measure the third particle axis length (i.e., depth / height; c axis). Aspect ratio measurements were grouped by length values into bins matching the Abakus measurements (i.e., 30 size bins; Sect. 2.2). We use the default factory settings, which have a predetermined calibration factor that converts pixels to micro-measurements (Fluid Imaging Technologies, 2011).

We checked the quality of DPI data output by assessing particle image clarity and particle length measurements using the edge gradient and particle length measurement metrics. Because Feret measurements use the particle image to identify particle length and width, images with poor clarity could introduce measurement errors. Therefore, we assess error within DPI measurements via image clarity, instrument limits, and previously published error assessment of particle aspect ratio measurements. More than 44 000 particles in total were measured throughout the sampling process with a minimum particle length of 2 µm and minimum estimated diameter of 1 µm (see Fig. S2 for particle counts by coarse particles of 5.1-6.4 µm as well as total counts and Fig. S3 for representative FlowCAM particle images from a single sample). The FlowCAM automated output generates three estimates of particle diameter based on (1) diameters based on the area of a circle, (2) mean value of each Feret measurement, and (3) weighted values of both methods 1 and 2 (Fluid Imaging Technologies, 2011).

We analyzed 54 samples via FlowCAM that were adjacent in depth to those measured on the CC (Fig. 1a). Blank measurements were collected at the start of each day and also between samples to ensure no background contamination; a total of 141 blank measurements were collected. Blanks were comprised of deionized water (DI) that was run for 10 min at a flow rate of $0.04-0.07 \text{ mLmin}^{-1}$. Prior to each run, the funnel and tubing were rinsed with a Citranox and DI solution, followed by a DI wash. Samples were then shaken and poured into the cleaned funnel and tubing. Samples were allowed to flow through a 0.5 mL cell prior to data collection to ensure that measurements were related to samples rather than the DI wash. The FlowCAM does not have the same size bin measurements as the Abakus and does not produce measurements for particle lengths of 2.1, 2.2, 2.4, and 2.7 µm. Because of the relatively similar mean widths and neighboring standard deviations, we used linear interpolation for these bin sizes to identify particle width (Figs. S4 and S5). Due to low particle counts $(5.1-6.4 \,\mu\text{m})$ in samples from 16-10 ka (Fig. S2), we focus our analysis hereafter on the 41 samples older than 16 ka (Fig. 1a). Although particle concentrations are high, we find no evidence of particle coincidence during HS1 or the LGM via our four checks of coincidence (-0.1 < r values < 0.1 and with p values < 0.01 and> 0.01; Table S1 in the Supplement). Following the methods and statistical tests for particle coincidence developed by Saey (1998) and used by Simonsen et al. (2018), the low rvalue and *p*-value statistics indicate that while there is a significant relationship between the ratio of coarse to fine particles and particle concentration, this explains very little of the observed methodological offset. Therefore, because there is no relationship between the ratio of coarse to fine particles and particle concentration, these results suggest that the overestimation of coarse particles in the PSDAbakus is related to assumptions of particle shape and sphericity rather than data collection processes. If the mismatch between the Abakus and CC was related to coincidence, then we would expect a significant relationship between the two parameters. Furthermore, we find that the deeper core sample (Heinrich Stadial 4 and 5) PSD mode values compared to Heinrich Stadial 1 and LGM mode values are statistically similar (Student's t test; p value > 0.05). Therefore, while there may be particle aggregation, it does not have a significant effect on the deep South Pole samples.

2.5 Abakus PSD calculation and calibration techniques

The main goal of this study is to explore the use of DPI and CC data to improve Abakus particle volume calculations. To this end, we used an array of methods to compare particle size metrics in Abakus data with concurrent CC and DPI samples. We compared four different methods for calculating particle volume and particle size distribution (PSD_{Abakus}) using a combination of Abakus, CC, and DPI data, each of which are briefly discussed below. The resulting particle volume, mass concentration, flux, and PSD_{Abakus} from each method are then compared to the corresponding CC particle volume, mass concentration, flux, and PSD_{CC}, which is assumed to be the most accurate method because the CC measures particle volume based on electrical impedance rather than through geometric calculations.

Aspect ratio data provide a direct test of the common assumption that ice core dust particles are spherical (i.e., Koffman et al., 2014; Ruth et al., 2003; Wegner et al., 2015) and allow us to calculate the influence of particle shape on a range of particle metrics. We apply our measured aspect ratios to the Abakus data using Eq. (1) (below). Because the FlowCam captures a two-dimensional snapshot of a threedimensional particle, two end-member calculations bracket the range of true particle volumes for a given aspect ratio (b/a; Fig. 2a and c). While we cannot measure the orientation of the particle in 3D space, the FlowCAM automatically measures the longest axis as the particle length (*a* axis) and the shortest axis as the particle width (*b* axis). Therefore, to account for particle height (*c* axis), we must assume that the height and/or depth of the particle is equal to either the *a* or *b* axis (Fig. 2). The assumption that the *c* axis is equal to the *a* or *b* axis results in volumetric changes (Fig. 2). By varying our height dimension, we can account for the missing particle dimension. We assume the following dimensions for each volumetric calculation: prolate assumes particle width (*b* axis) = height (*c* axis; Fig. 2a), oblate assumes particle length (*a* axis) = height (*c* axis; Fig. 2b), and sphere assumes all lengths equal to the particle length (*a* axis; Fig. 2c).

Ellipsoidal 1 Volume =
$$\frac{4}{3}\pi(ab^2)$$
 (1a)

Ellipsoidal 2 Volume =
$$\frac{4}{3}\pi(a^2b)$$
 (1b)

Spherical Volume =
$$\frac{4}{3}\pi r^3$$
 (1c)

In all cases, DPI measurements have been binned by particle length under the assumption that length measurements of particles are equal between the Abakus and DPI. While we recognize that the Abakus may be recording the extinction cross-section rather than true particle length, we assume that both measurement techniques (Abakus and DPI) represent accurate particle length since the Abakus was calibrated to CC measurements (Koffman et al., 2014). Based on the dimension assumptions above, prolate represents a minimum volume, oblate represents a moderate volume, and sphere represents the null hypothesis of equal particle dimensions (maximum volume). Thus, three different calculation methods, corresponding to Eqs. (1a), (1b), and (1c), are applied to generate three distinct datasets from the Abakus measurements, each with a corresponding volume, mass concentration, flux, and particle size distribution. For spherical volume, r = 1/2 particle diameter. For ellipsoidal volume, a = 1/2 particle length, b = 1/2 particle width, and the c axis (particle height) is set to equal particle width (b axis; prolate) or particle length (a axis; oblate).

Adjusted Abakus size bins =
$$\sqrt[3]{\overline{x}} \times \text{Abakus size bins}$$
 (2)

 $A_N = c_i \times \text{Abakus size bins}$ (3a)

Adjusted Abakus size bins

$$= \min\left(\left|\ln(A_{\rm N}) - \ln(\rm CC_{bins})\right|\right) \tag{3b}$$

In addition to the direct comparisons of particle metrics, we also explore methods that statistically adjust the Abakus data to match the more accurate CC data. We use two techniques developed by Simonsen et al. (2018) that were specifically designed to bring PSD_{Abakus} in line with PSD_{CC} . The calibration techniques shift PSD_{Abakus} using a spherical particle volume by corresponding averaged size aspect ratios and corresponding PSD_{CC} (described below). In the first calibration method, we multiply the cubic root of the median

aspect ratio for bin sizes $2.0-6.4\,\mu\text{m}$ by the original Abakus bin sizes (Eq. 2), generating a new set of PSD_{Abakus} values. The second technique seeks to stretch, contract, and/or shift the calculated PSD_{Abakus} to more closely match PSD_{CC} following methods outlined by Simonsen et al. (2018). Here, we multiply values from the PSD_{Abakus} by scalars ranging from 0.55-1.10 and then optimize the match with the PSD_{CC} by minimizing the difference in the absolute value of the natural log of PSD_{Abakus} and PSD_{CC} (Eq. 3). In Eq. (3), c_i is equal to all values ranging from 0.55–1.10 in increments of 0.01, A_N represents linearly scaled Abakus size bins, and CC_{bins} represents Coulter counter size bins. The applied shift results in the calibrated PSD_{Abakus}. While our study focuses on the change in shape based on 2D particle dimensions, we discuss the comparisons to both calibration techniques (Eqs. 2, 3a. 3b) in Sect. 4.3.

We used a combination of the above techniques (Eqs. 1, 2, 3a, 3b) to generate seven sets of Abakus data and their associated metrics. For clarity, we categorize our analyses into calculations (Eqs. 1a-1c) and calibrations (Eqs. 2, 3a, 3b); calculations refer to simple geometric differences (i.e., spherical vs. ellipsoidal volume differences), and calibrations refer to shifts in Abakus data relative to CC data. Equation (1) calculates particle volume using three different particle axis dimensions. Equation (2) shifts the PSDAbakus values based on aspect ratio measurements. We do not use Eq. (2) for ellipsoid PSD_{Abakus} values because they already incorporate aspect ratios into their calculations. Finally, Eq. (3a) and (3b) manually shifts the PSDAbakus values to best fit PSDCC. Figure 2 shows a depiction of the different hypothesized particle shapes and the resulting volume differences from changes in their geometries.

Total offset =
$$\sum_{i=\text{size bin} \ge 2.0}^{6.4} |(\text{Abakus}_i/\text{CC}_i) - 1|$$
(4)

To summarize the difference between the methodologies, we use Eq. (4), which we refer to as "total offset". Here, we assume that the particle sizes measured by the CC are the best representation of the true particle sizes in the ice core. Conceptually, the total offset metric provides a summary statistic integrating the mismatch between Abakus and CC data across bins. This allows us to assess the various methodologies we apply to correct the Abakus output for non-spheroidal particle geometries. The total offset compares the ratio of the dvolume/dln(diameter) Abakus to the CC, the metric used to assess particle size distributions in ice cores following Ruth et al. (2003). A value of 1 is subtracted from the ratio of Abakus to CC to shift an equal value to 0. The total offset, calculated for Abakus data, measures the sum of the differences from a 1:1 ratio between Abakus and CC PSDs using Abakus-defined bin sizes. The Abakus_i /CC_i describes the Abakus median calculation or calibration per size bin relative to the median CC sizes of similar size bins. In summary, based on this calculation, the lower the total offset the closer the two parameters are to each other.



Figure 2. Representative ellipsoidal and spherical particles with labeled axes showing (**a**) prolate ellipsoidal particle shape with a FlowCAM image of an SPC14 particle, (**b**) oblate particle with rotated image of the particle from panel (**a**) (because particle images are in 2D we cannot differentiate between oblate and prolate), and (**c**) spherical particle shape with a FlowCAM image of an SPC14 particle. Left-hand images are representations of hypothesized axis measurements and geometries. For each particle, the assumptions and geometric volume calculations are provided. The volume calculations highlight the large difference in particle volume between equal and unequal axes (Eq. 1).

Table 1. Total offset measures for each calibration methodology relative to PSD_{CC} values. Total offset is minimized for prolate CC calibration, making it the preferred approach for this dataset.

	HS1	Glacial	HS4 and 5	Total offset
Sphere PSD calculation	38.82	22.16	13.02	73.99
Prolate PSD calculation	9.11	4.49	4.81	18.41
Oblate PSD calculation	19.02	8.64	4.56	32.22
Aspect ratio	33.15	19.26	11.51	63.92
Abakus spherical CC calibration	9.21	22.03	12.12	43.36
Abakus prolate CC calibration	1.53	1.63	6.64	9.80
Abakus oblate CC calibration	2.31	8.89	19.23	30.43

3 Results

3.1 Abakus and Coulter counter measurement relationships

Abakus and CC mass concentrations are positively and significantly correlated (Fig. S6; r = 0.95, p value < 0.001, lin-

ear slope_{mass} = 2.32), consistent with the findings of Ruth et al. (2008, i.e., $R_{log} = 1.00$). However, the Abakus mass concentration (using the assumption of spherical shape) is nearly twofold greater (~ 2.3) than that derived from the CC (Fig. S6), and the average PSD_{Abakus} mode values for all 41 samples is $4.80 \pm 2.42 \,\mu\text{m} (2\sigma)$, while mean PSD_{CC} is $2.70 \pm 2.73 \,\mu\text{m} (2\sigma)$, suggesting that the Abakus overestimates particle mass (i.e., volume). Abakus to CC PSD ratios (Figs. S6 and S7) show that while finer particles ($< 3.0 \,\mu m$) are closer to a 1:1 scale, coarser particles $(3.0-5.0 \,\mu\text{m})$ are elevated (Abakus: CC > 2), which indicates that coarse particles account for the bulk of the methodological offset (Fig. S7). This nonlinear relationship between fine and coarse particles measured in the Abakus and CC demonstrates that methodological variability needs to be accounted for. Furthermore, the distribution differences between PSD_{Abakus} (under spherical shape assumption) and PSD_{CC} are not temporally consistent (Figs. 3, S7 and S8), potentially indicating changes in particle shape over time. Specifically, PSD_{Abakus} and PSD_{CC} are most similar during HS4 and 5 (total offset 13.02; Table 1; Fig. 3c and f), followed by the LGM (total offset 22.16; Table 1; Fig. 3b and e) and HS1 (total offset 38.82; Table 1; Fig. 3a and d). Across all time periods (41 samples representing 34 kyr), the Abakus measures greater numbers of coarse particles relative to the CC, with the largest overestimation during HS1 (Fig. 3a and d). While particle sizes of 3-5 µm have the highest offset, the finest particles $< 2.0 \,\mu\text{m}$ throughout all time periods have similar apparent PSD ratios produced by Abakus and CC (Fig. 3a-f).

3.2 Dynamic particle imaging

ratios All size-dependent aspect (width / length =b axis /a axis) have median values below 0.95, indicating that the majority of particles are ellipsoidal (elongated) and not spherical (Figs. 3a, c, and S10). Size-dependent aspect ratio measurements are defined by two groups: (1) finer aspect ratios ($< 5 \mu m$) and (2) coarser aspect ratios $(\geq 5.1 \,\mu\text{m}; \text{ Fig. S10})$. Fine aspect ratios have a narrow distribution, while coarser aspect ratios have a broader distribution and are more elongated (Figs. S10 and S11). Fine particles have median aspect ratios of $0.76 \pm 0.10 (2\sigma)$. Coarse particles have a median aspect ratio of 0.70 ± 0.14 (2σ) . Across all 41 samples, distributions of particle aspect ratio have a left skew towards elongated particles and are leptokurtic (Figs. S10 and S11), indicating that along with a left skew, particles are more likely to be asymmetrical rather than symmetrical (Fig. 4a and c). There is no relationship between particle size bins and distribution skewness (Fig. 4a and c). While Mathaes et al. (2020) provided evidence of high error ranges in aspect ratio measurements using DPI at $10 \times$ for particles $< 5 \,\mu$ m, our aspect ratio measurements, obtained using a greater $20 \times \text{zoom}$ (Fig. 4b and d), had low and consistent aspect ratio standard deviations.

There are different temporal trends in the fine and coarse particle aspect ratios. Coarse particles are more variable through time and are slightly more elongated than fine particles (Fig. 4b and d). Across the three time periods studied, we found coarse particle aspect ratios of 0.68 ± 0.12 (median $\pm 2\sigma$, HS1), 0.70 ± 0.09 (LGM), and 0.71 ± 0.08 (HS4 and 5). In comparison, fine particle aspect rates were 0.76 ± 0.07 (HS1), 0.76 ± 0.02 (LGM), and 0.77 ± 0.05 (HS4

and 5). Using a *t* test, we found that coarse and fine particle aspect ratios were significantly different for each period (*p* value < 0.05). The variability of both fine and coarse particle aspect ratios appears to be higher during HS1 compared to the LGM and HS4 and 5, though sample size may have contributed to these differences (HS1: n = 6, LGM: n = 19, and HS4 and 5: n = 16). Because of these differences in distribution and the nonlinear offset between PSD_{Abakus} and PSD_{CC}, we explore the differences between fine and coarse particles. For clarity, we refer to PSD_{Abakus} and PSD_{CC} when referring to the entire distribution of particles by size and fine and coarse particles when referencing general trends based on distribution differences in aspect ratios.

Width measurements (equal to the bin size multiplied by the FlowCam aspect ratio) afford the opportunity to calculate non-spherical particle volume more accurately than under the assumption of equal width and height (prolate and oblate; Fig. 2a and b). Width measurements ranged from 1.18–4.3 µm. We use linear interpolation for width measurements in bins 2.1, 2.2, 2.4, and 2.7 µm because of similarities in averages and 2σ for neighboring bin sizes (Figs. S4) and S5). Width measurements are not statistically different between the three time periods (Student's t test, p value >0.1). Because of the temporally consistent average particle width measurements (HS1, 2.33 ± 1.87 ; LGM, 2.45 ± 2.07 ; HS4 and 5, 2.40 ± 2.03 ; 2σ) for each bin in the SPC14 samples, we use average width measurements in particle volume calculations for the entire core and all dust particle metrics (Eq. 1; Figs. S4 and S5). Therefore, we are assuming that changes in aspect ratio are primarily driven by variations in particle length.

Abakus calculation technique comparison (Eqs. 1a–1c)

Following suggestions by Simonsen et al. (2018) we utilize the DPI results and evaluate two different metrics for calculating Abakus data. Calculated metrics include (1) spherical volume estimates (i.e., null hypothesis, Eq. 1c) and (2) ellipsoidal volume estimates (i.e., elongated shape hypothesis, Eqs. 1a and 1b). During all time periods, the spherical PSD_{Abakus} overestimates all particle sizes, having the greatest effect on the coarsest particles (Fig. 3a-f and Table 1). During HS4 and 5, oblate PSD calculation has the lowest total particle offset (closest to the CC) of all calculation and calibration techniques, which is followed by prolate PSD calculation (Table 1). Coarse particle PSD_{Abakus} values calculated using either prolate or oblate (Eq. 1) volumes are closer to PSD_{CC} values in all time periods compared to average spherical volumes (Fig. 4b and d). Fine particle shape calculations do not have a consistent relationship with CC values. During HS1, prolate particle shapes closely match CC values, while during the LGM and HS4 and 5, oblate particle shapes more closely match PSD_{CC} values. Ellipsoid metrics produce the lowest total offset values for all time periods (Ta-



Figure 3. Comparison of Abakus volume and calibration schemes (Eqs. 1, 3a, 3b, 3c) to Coulter counter data (**a**–**c**) and the ratio of particle size distributions between Abakus and Coulter counter data (**d**–**f**). Sphere: Abakus sphere PSD calculation (length = width = height), CC: CC PSD calculation (accurate volume), prolate: Abakus prolate PSD calculation (width = height), oblate: Abakus oblate PSD calculation (length = height). The top panel compares the PSD_{Abakus} volume and calibration techniques to PSD_{CC}, and the bottom panel is the ratio of each PSD_{Abakus} to the PSD_{CC} in different time periods. The dotted line in the lower panels represents a 1 : 1 value, and the closer each colored line is to the 1 : 1 line, the lower the total offset. Using ellipsoidal volume calculations always improves methodological correspondence, although the type of ellipsoid that best fits the data depends on particle size. For a comparison of all calculations and calibrations see Fig. S9.

ble 1). Of all three calculation metrics, prolate PSD calculations have the lowest total offset between HS1, LGM, and HS4 and 5, followed by oblate PSD. Ellipsoid calculations (both prolate and oblate) reduce total offset relative to spherical shape calculations by between ~ 25 % and 44 % (Table 1; Fig. 3a–f).

4 Discussion

Calculating Abakus particle volume metrics using ellipsoid volumes reduces the discrepancy (offset) between Abakus and CC methodologies. Based on our results, we suggest there are three significant implications for Antarctic ice core microparticle analyses: (1) average particle shape is consistently ellipsoidal, not spherical, which likely represents microparticle mineralogy (e.g., clay particles and other sheet silicates); (2) assuming spherical particle shape leads to overestimation of coarse particle volume and/or mass; (3) using DPI, particle width measurements can be used to reduce observed offsets between Abakus and CC data; and (4) changes in particle shape through time (or a lack thereof) can be used as an additional piece of information to assess changes in the dust cycle. The closer correspondence between ellipsoid values and PSD_{CC} effectively reduces the magnitude of the Abakus offset that occurs when assuming a spherical particle shape (Fig. 5). Younger than 16 ka, there is a discrepancy between the Abakus and CC values. We discuss the discrepancy in Sect. 4.1 and the three implications of our microparticle analyses in Sect. 4.2. While using ellipsoid volumes is important for reducing discrepancies among different microparticle measurement techniques, it also supports conclusions of previous analyses of microparticle mineralogical composition and creates new avenues for future research, which are discussed in Sect. 4.5.

4.1 Effect of Estisol-140

During the last glacial period until 16 ka, there is a close correspondence between all metrics (mean and median resampled data) from the Abakus and the CC (Figs. 5 and S12). However, Estisol-140 affects both the Abakus and CC data from 16–0 ka because of the relatively low dust signal. The clearest example of this is a stepwise increase in particle concentration with the introduction of the drilling fluid at 160 m (~ 2 ka; not shown). Because dust signals during the interval from 16–10 ka are 3–5 orders of magnitude lower than dur-



Figure 4. Particle aspect ratio data as a function of aspect ratio (**a**, **c**) and time (**b**, **d**). (**a**, **c**) Aspect ratio distributions (medians) for HS1, LGM, and HS4 and 5. (**b**, **d**) Median aspect ratio values (colors are related to time in panels **a** and **c**, respectively) with 2σ error bars plotted with the SPC14 δ^{18} O record (Steig et al., 2021; dark grey), SPC14 particle concentration record (brown), and WAIS Divide CO₂ (Bauska et al., 2021; Marcott et al., 2014; salmon). The top two panels (**a**, **b**) show coarse particle (> 5.0 µm) data, and panels (**c**) and (**d**) show fine particles (< 5.0 µm). The variability (2σ) increases in samples older than 18 ka. Median aspect ratio values are identified by the black dashed line. Results show that aspect ratios are skewed towards more elongate values (< 1) with fine and coarse aspect ratios of 0.76 ± 0.5 and 0.70 ± 0.10 (2σ), respectively.



Figure 5. Comparison of CC and Abakus spherical and ellipsoidal (prolate, Eq. 1b) particle data including (**a**) volume mode, (**b**) flux, (**c**) mass concentration, and (**d**) number concentration. The Abakus-calculated prolate data are a better fit to the CC data for all volume-related metrics (flux, mass concentration, ppb, and PSD). Number concentrations (**d**) are unaffected by particle shape.

ing the LGM, the influence of Estisol-140 rivals the true microparticle signal in the younger samples. While we are able to remove contamination peaks caused by drilling fluid in the continuous Abakus data, we are not able to separate these signals in the lower-resolution CC samples during times of low dust concentrations. Therefore, we do not interpret CC samples younger than 16 ka.

4.2 Ellipsoid particle shape

Our measurements (Figs. 3-5 and Table 1) show that average particle shape is ellipsoidal, consistent with previous studies (Simonsen et al., 2018; Potenza et al., 2016). In SPC14, coarse and fine particle shapes are distributed around median aspect ratio values of 0.70 and 0.76, respectively. Using an ellipsoidal particle shape informed by PSD_{Abakus}: PSD_{CC} measurements reconciles most of the total offset between Abakus and CC data (Table 1; Fig. 5). Together with the FlowCAM images, these results confirm that microparticles are predominantly elongated rather than spherical. Prolate (Fig. 2a) geometries have shorter width / height measurements compared to oblate or spherical geometries. Our assumption of consistent particle width measurements seems to be supported by our results (i.e., reduced discrepancy between Abakus and CC; Figs. 3-5 and Table 1). Future studies should use this assumption with caution after measuring sitespecific particle metrics on the timescale in question. Mathaes et al. (2020) determined that measurement techniques can lead to greater variability of 5 µm spherical (aspect ratio 1.0) or elongated particles (aspect ratio 0.2). However, our aspect ratio measurements consistently cluster between values of 0.70 and 0.76 across a wide range of time periods and particle sizes (Fig. S10). Not only are these aspect ratios outside the range of spherical and elongated particles (as defined by Mathaes et al., 2020), but their consistency demonstrates the repeatability of our techniques. Despite our millennialscale consistency in aspect ratios at the South Pole, Potenza et al. (2016) determined that during short decadal time periods (i.e., ~ 15 years), particle aspect ratio is variable, indicating that high-frequency changes in particle shape may exist despite the low-frequency stability. Furthermore, it is important to note that our samples occur during 50-16 ka, prior to interpreted atmospheric reorganization and changes in dust particle source area variability (Wegner et al., 2012; Aarons et al., 2017). Therefore, while our results under the assumption of assumed constant particle width measurements do reduce discrepancies between Abakus and CC volume-related metrics, we advise caution when using this assumption without measuring particle shape at different temporal scales and spatial locations.

While most particles are consistently ellipsoidal, ellipsoid types (Fig. 2a and b) are not consistent through time. During HS4 and 5, oblate (Eq. 1c) provides a better fit to the CC data than prolate PSD (Eq. 1b), which has a better fit during HS1 and the LGM. While aspect ratio measurements remain relatively constant through time, it is likely that the third, hidden dimension of particle height (axis c) varies through time. PSD_{CC} values lie between prolate and oblate PSD_{Abakus} values, indicating that our two ellipsoidal geometries represent bounds on volume for a given aspect ratio. This information could imply a shift in composition from prolate–linear to oblate–planar shaped particles between 36 and 27 ka, perhaps indicative of a change in source environment or mineralogy.

While a study of Saharan dust transport with a larger range of sizes (i.e., Von Holdt et al., 2021) identified particle roundness ranging from 0.2–1.0 (aspect ratio) with varying complex mineralogy, our study shows relatively consistent and invariant fine particles $< 5.0 \,\mu\text{m}$ with slightly more variable coarse particles. The consistency of South Pole particle aspect ratios may be indicative of the distal continental source during the last glacial period, which we infer to be South America based on geochemical analyses of dust from Vostok and Dome C (Delmonte et al., 2008, 2004). The slight temporal variability in the coarser particles may indicate some mineralogical and/or source variability (e.g., Arnold et al., 1998).

Particle mineralogy and shape in ice cores are mechanistically linked. Arnold et al. (1998) determined that particles that are transported farther are smaller and tend to be composed of more platy to secondary minerals (i.e., clays). Particles that were transported from closer dust sources tended to be coarser in size and composed of more primary minerals (i.e., plagioclase). This relationship was utilized by Paleari et al. (2019) to identify variations in source region between the Holocene and the LGM at Dome C, Antarctica. Paleari et al. (2019) noted that although LGM and Holocene particles were mineralogically similar, the Holocene contained more volcanic and metamorphic minerals, as well as those indicative of strong weathering. Although they identified some mineralogical variability, they interpreted these results as an indication that there had not been a significant shift in dust source region to Dome C between the LGM and Holocene. Prior research on the mineralogy of modern dust at the South Pole indicates that particles are predominantly composed of clay minerals, specifically illite (20%), kaolinite (8%), halloysite (4%), vermiculite (3%), and other related clay minerals (24%; Kumai, 1976). The relatively small grain sizes coupled with predominant clay mineralogy at the South Pole is indicative of long-range transport. Recent studies have also identified elongate diatoms in snow and ice samples from Dome C and western Antarctica, providing further evidence of the wide range of particle shapes in Antarctic aeolian samples (i.e., Delmonte et al., 2017; Allen et al., 2020). While small clay particles ($\sim 2.5 \,\mu m$) are predominantly oblate particle shapes, larger particles ($\sim 10 \,\mu m$) became more variable in shape (Meland et al., 2012). The close correspondence between our PSD_{Abakus} and PSD_{CC} fine particles supports their interpretation. The disagreement between our PSDAbakus and PSD_{CC} coarse particles further suggests that coarse particles are more variable in shape.

Particle shape can affect particle deposition and radiative properties (i.e., Li and Osada, 2007a, b; Von Holdt et al., 2021, and references within; Knippertz and Stuut, 2014). For instance, rounder particles tend to settle out first, while elongated particles remain in the atmosphere longer due to resistive forces (Ginoux, 2003; Formenti et al., 2011, and references within). The gravitational center location on individual dust particles, particularly if asymmetric, can also influence settling and transport (Li and Osada, 2007b). While particle shape can have negligible effects on light scattering, it can impact aerosol optical depth calculations (a technique used to calculate atmospheric dust concentrations) by 30 %– 40 % (Potenza et al., 2016; Feng et al., 2015).

While our data suggest that particle width is constant throughout our record, we acknowledge that this may be due to our relatively low resolution of our analysis. For example, Potenza et al. (2016) measured aspect ratios of particles recovered from EPICA Dome C, which varied between prolate and oblate within 10–15 years. Simonsen et al. (2018) also state in their interpretations that more extreme aspect ratios are more likely related to long-distance transport, during which short dust storms increase variability. Because we are measuring selected millennial-scale aspect ratio variability, this suggests that our interpretations could reflect constant average dust source weathering and transport conditions to the South Pole, since particle shape can influence settling and transport (Li and Osada, 2007b, and references within; Ginoux, 2003; Formenti et al., 2011). We suspect that analyses of short-term variability would likely highlight more variability in particle width.

4.3 Particle size and shape effect on calculated volume

Our results indicate that the accuracy of certain Abakus data metrics can be improved by incorporating measurements of particle shape (i.e., aspect ratio). Incorporating aspect ratios reduced the total offsets by 25 %-44 % (Table 1). While any volume-related Abakus particle metric requires calculation and possibly calibration using the DPI and CC techniques, the non-volume-related metrics (i.e., particle number concentration) do not require calibrating to external methods. Particle number concentration showed no evidence, even during periods of high concentration, of coincident counts. Our tests for particle coincidence indicate that there is either a weak but significant relationship (p value < 0.01) or no relationship between the ratio of coarse to fine particles and the particle concentration. Furthermore, the Abakus concentrations scaled to CC sample resolution (i.e., 54 discrete samples) and CC samples have a significantly strong correlation (r = 0.95, p value < 0.01; Fig. S6), similar to relationships published in Ruth et al. (2008). Our results indicate that number concentration measurements derived from highresolution particle Abakus measurements are accurate, in agreement with previous microparticle methodological analyses (Ruth et al., 2008).

By using the ellipsoid volume calculations alone, differences between the Abakus and CC flux and mass concentration were reduced within each time period (Figs. 3 and 5b, c). Ellipsoidal volumes reduced Abakus median particle flux and mass concentration relative to the corresponding CC values from a ratio of 1.82 (with an assumed spherical volume) to 0.79 (with an assumed prolate volume) and 1.20 (with an assumed oblate volume, where a value of 1 is equal). Without calculation and/or calibration using DPI and CC techniques, assumptions made while using Abakus volume metrics could overestimate mass and flux up to $\sim 980\%$ (i.e., HS1; Fig. 3d). In agreement with previous studies (Simonsen et al., 2018; Potenza et al., 2016), we find that SPC14 particles are elongated rather than spherical, fitting somewhere between oblate and prolate particles (Figs. 2 and 3; Potenza et al., 2016; Simonsen et al., 2018). Using prolate volumes compared to spherical volumes reduced the average percent difference between the Abakus and CC by 221 %, 147 %, and 105 % during HS1, the LGM, and HS4 and 5, respectively. Using oblate (oblate) volumes reduced the average percent difference between Abakus and CC metrics by 130 %, 89 %, and 66 % during HS1, the LGM, and HS4 and 5, respectively. Percent difference values that are over 100 % represent overcorrection of the calculation and/or calibration method.

Particle mode is a useful measurement for analysis of varying PSD and has been used to identify changes in atmospheric structure (i.e., Koffman et al., 2014; Delmonte et al., 2017; Ruth et al., 2003). Although our calculation techniques did lower mode particle diameter values overall, they are still within the 2σ envelopes of the original Abakus spherical PSD calculation $(4.80 \pm 2.42, 3.20 \pm 2.41,$ 3.20 ± 2.60 ; spherical PSD, prolate PSD, and oblate PSD mode values; Fig. 5d). And although the median (average) PSDAbakus mode is reduced overall and closer to the CC PSD mode value $(2.70 \pm 2.74; 2\sigma)$, a t test between the four metrics shows that spherical and oblate PSD are significantly similar (p value > 0.05), while prolate and CC volumes are statistically different (p value < 0.05). While the standard deviation in mode values is high, these results suggest that prolate mode PSD values are the most similar to CC values.

4.4 DPI as a technique for particle calibration

DPI improved Abakus particle volume metric calculations and calibrations by providing width and aspect ratio measurements that refine calculations of particle volume. Our results highlight the fact that spherical particle shape assumptions are not only incorrect, but their calibration techniques at the South Pole were also less effective compared to calculation and/or calibration techniques that used ellipsoidal shapes and volumes. Total offset values and coarse particle offset clearly indicate that the use of width measurements from DPI improved the fit of Abakus to CC data. Given sufficient DPI measurements to characterize any temporal or size-dependent variability in particle shape, this technique could be used to calibrate Abakus volume, mass, and flux estimates in future ice coring studies. While DPI does not provide 3D measurements, the assumption of varying height to either particle length (oblate) or width (prolate) still resulted in decreased discrepancy between the Abakus and CC (i.e., reduced total offset; Table 1). Measurements of particle shape also greatly improve understanding of particle size distribution in Abakus data. This nonlinear offset between Abakus and CC could lead to potentially spurious interpretations of mode particle size, a commonly used metric for interpreting transport distance and wind strength (Koffman et al., 2014; Ruth et al., 2003; Delmonte et al., 2017; Steffensen, 1997). Use of DPI with previously published calibration techniques further reduces Abakus volume-related metrics compared to similar CC data (Simonsen et al., 2018; Potenza et al., 2016). We also note that the reduced discrepancy between the Abakus and CC data supports our hypothesis that DPI particle geometries can be matched with Abakus particle length (extinction cross-section) measurements. If the Abakus geometric dimensions were extremely biased on millennial timescales, then we would expect the use of DPI not to reduce discrepancy between the two datasets.

We recommend the following steps for the use of DPI as a technique for more accurate volume-corrected Abakus particle metric calibration: (1) select samples for DPI based on changes in Abakus concentration, flux, and/or PSD; (2) adjust particle measurements to match Abakus bin sizes; (3) vary measurements to account for 3D variability (i.e., height value) between prolate, oblate, and spherical particle shapes; (4) compare new calculated particle size distributions using the total offset parameter as a metric for comparison between Abakus and CC data; and (5) use the particle shape calculation that produces the lowest total offset between the Abakus and the CC.

4.5 Implications of South Pole particle shape in paleoclimate reconstructions

Temporal variations in particle shape may provide additional information on past climate, atmospheric, and environmental dynamics because transport and depositional velocities, dust mineralogy, and transport distance all likely have an influence on particle shape at a particular depositional location through time (Von Holdt et al., 2021; Knippertz and Stuut, 2014; Li and Osada, 2007b). Each of the time periods highlighted in this study (HS1; LGM; HS4 and HS5) represent times of significant climate change based on the SPC14 δ^{18} O record (e.g., temperature) and atmospheric CO₂ variability (Figs. 1 and 3). Dust transported into interior East Antarctica is predominantly from southern South America via the Southern Hemisphere westerly winds (Delmonte et al., 2008, 2020). Because of the consistency of particle shape and similarities between the Abakus volume-corrected particle flux and mass metrics with similar measurements made on the CC, the calibrated SPC14 particle record can be used as a robust volume-accurate metric for atmospheric dynamics over the past 54000 years. The consistency in particle shape further implies consistent mechanisms of transport.

5 Conclusion

We used dynamic particle imaging (DPI) to assess microparticle shape in the SPC14 record between 50 and 16 ka. Our results represent the longest and highest-resolution record of particle shape yet developed from ice cores. Through DPI we were able to (1) use average particle dimensions to lower inaccuracies associated with shape assumptions, (2) generate new techniques for calculating particle metrics from Abakus data, and (3) develop a new calibrated SPC14 Abakus dust dataset. Our comparison of SPC14 microparticle data from three different instruments and methodologies reveals consistently ellipsoid particle shapes, though with some temporal variations in inferred axis rotation (e.g., prolate vs. oblate). In agreement with findings from the Renland Ice Cap (Greenland; Simonsen et al., 2018), our results demonstrate that microparticles deposited in polar snow and ice are consistently oblong and not spherical. Assumptions of sphericity lead to larger offsets between Abakus and CC data, greatly increasing volume-based metrics in the Abakus data. Our recommendations include the following steps for calculating the best fit for Abakus-derived volumetric data: (1) use DPI to obtain length and width measurements for particles; (2) calculate ellipsoidal volumes and PSD_{Abakus}, and use Simonsen et al. (2018) method 3 for adjustments, depending on the application; and (3) minimize the total offset parameter. We recommend for future dust volume calculations that a robust reconnaissance of dust particle shape be completed based on drill site and dust profile characteristics (i.e., concentration, size, chemistry, and/or source region variability). The methods presented here produce a high-resolution continuous Abakus record that is most comparable to discrete CC measurements. The consistency in particle shape suggests that consistent transport mechanisms occurred between 50 and 16 ka.

Code availability. The Python-based code used to clean the SPC14 Abakus dataset is available at https://github.com/ katherine-anderson/SPICEcore_Dust_Data_Processing (last access: 9 November 2020; Anderson, 2020).

Data availability. Dust datasets used in this paper are available from the US Antarctic Program Data Center (https://doi.org/10.15784/601553; Kreutz, 2022).

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Author contributions. AC collected and analyzed the data and prepared the paper with contributions from all co-authors. KK, EO, JCD, and MW designed the experiments and guided the overall research effort. DF and ZT melted the core, measured the Abakus data, and collected discrete samples. BK supported and provided the laboratory for Coulter counter measurements. JM supported Flow-CAM measurements. KA wrote the Abakus cleaning code. DW provided data analysis and writing support. NH measured a subset of the Coulter counter samples. AP provided writing support. MH provided laboratory analysis feedback.

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