1 VolcAshDB – A global database of volcanic ash particles

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- 19 Scientific Data
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21 Abstract (Maximum 170 words recommended)

22 Volcanic ash is made of particles smaller than 2 mm and is produced during volcanic eruptions. 23 Studying the properties of volcanic ash is key for a range of applications, including volcano monitoring. 24 However, given the large variety of textures, colors, and shapes of particles from different eruptions 25 and volcanoes it is challenging to classify them in a reproducible and systematic manner. Here, we 26 present an open access web-based platform that displays the contents of a database of volcanic ash 27 particles, VolcAshDB (https://volcashdb.ipgp.fr/). Particle data includes optical microscope images, 28 physical characteristics of the shape, texture and color, and a classification label with the particle type 29 (free-crystal, altered material, juvenile or lithic). It currently hosts data of 12,044 particles from 53 30 samples across 13 volcanoes that can be visualized and downloaded in a user-friendly manner. The 31 database has already been used to develop particle classifiers with Machine Learning. We plan to 32 continue it to grow and encourage the volcano community to contribute towards standardizing 33 volcanic ash classification.

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35 Background & Summary

36 Volcanic ash is made of particles smaller than 2 mm in diameter and is produced during explosive 37 eruptions. The study of its chemical composition and physical properties is important for a range of 38 applications, from understanding the impact of volcanic plumes to the Earth's climate¹ or air traffic², 39 through human health impacts³ as well as to monitoring the beginning and end of an eruption, and transition among eruptive styles^{4–7}. In particular for monitoring purposes, using volcanic ash requires 40 41 the proper identification and characterization of its particles into different components (free crystals, 42 altered material, lithic, and juvenile). However, it is commonly difficult to unequivocally associate the 43 features observed under the optical and scanning electron microscopes to a given particle type⁸. 44 Moreover, the criteria for classification and measurement of particle characteristics are not 45 standardized, and thus, the volcanic ash particle data are hard to compare between different 46 eruptions and volcanoes.

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To alleviate this situation, Benet et al.⁹ created the Volcanic Ash DataBase (VolcAshDB) which aims at 48 49 compiling the physical and visual characteristics of ash particles from a range of eruptions and 50 volcanoes using a standardized methodology. As of now, data of 12,044 ash particles from 25 51 eruptions across 13 volcanoes (Figure 1A) have been collected. The dataset was used to develop 52 Machine Learning (ML)-based models that are able to automatically classify volcanic ash particles into 53 different categories¹⁰ and thus opens opportunities for comparative studies between eruptions and 54 for time-series analysis that could be used during volcanic crises. Here, we present the database 55 design, web interface (https://volcashdb.ipgp.fr), and resources to download, navigate and visualize 56 the database contents. Our plan is to continue to grow the database to increase its robustness and 57 representativity of the wide ranges of volcanic ash, and to encourage the volcano community to 58 contribute to VolcAshDB to advance towards standardising volcanic ash particle classification. 59

60 Methods

61 Procedure for image acquisition of ash particles. Samples were first ultrasonically cleaned, dried and 62 sieved using four meshes of 0¢ (1 mm), 1¢ (0.5 mm), 2¢ (0.25 mm), and 3¢ (0.125 mm) pore-size 63 dimensions. Using the coarsest available grain-size per sample, particles were arranged on adhesive-64 coated glass slides for scanning on a binocular scanning stage. Several scans at different focal depths 65 were obtained and "fused" into a 2D-array with a good particle focus. The resulting scans were 66 segmented and color normalized to generate individual, high-resolution (~1,800 pixels/mm in 67 average) particle images that are the main data type of the dataset (see more details in Benet et al.⁹). 68 The images were obtained using a Leica binocular microscope equipped with an LMT260 XY Scanning 69 Stage and a LAS X imaging software at Nanyang Technological University, Singapore (Benet et al.⁹), 70 and a binocular scanning stage, Keyence VHX-7000, coupled with its Keyence machine system and 71 software at Institut de Physique du Globe de Paris (IPGP).

73 Extraction of features from images of particles. Particle images were quantitatively analyzed to 74 extract 33 measurements of particle properties (Figure 1B), hereafter called features. These 75 characterize the particle shape, texture, and color (see Figure 1C for three examples; and Benet et al.⁹ 76 for their definition and calculation). Shape features were obtained from the particle contour and have 77 been extensively used to characterize perimeter-based irregularities, particle-scale cavities, and/or overall external morphology^{11–13}. Textural features were quantified from the distribution of grayscale 78 79 pixel intensity values on the particle surface. Features such as the dissimilarity or correlation are 80 calculated from the Gray Level Co-occurrence Matrix (GLCM¹⁴), to distinguish for example between high textural complexity and uniform smoothness. Color features were extracted as the mean, mode 81 82 and standard deviation of the distribution of each channel in both the Red-Green-Blue (RGB) and Hue-83 Saturation-Value (HSV) color spaces. These features are sensitive to chromaticity, indicative of 84 dominant color hues, intensity, and brightness. The procedure for feature extraction was automated 85 through a Python script available in a public GitHub repository (https://github.com/dbenet-86 max/volcashdb_classification) and run in the High-Performance Computing center S-CAPAD at IPGP.

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88 **Procedure for image classification.** Ash particles were classified from visual inspection of the images 89 into free-crystal, altered material, juvenile and lithic types (Figure 1C) following the criteria described 90 in⁹. The main approach involves recognizing diagnostic features for each type. In short: Free crystals 91 are identified by their well-faceted shapes, sometimes exhibiting twinning and cleavage, and color. 92 Altered material appears in colors ranging from white to yellowish, with a granular texture and a 93 variety of hydrothermal secondary minerals, sometimes showing devitrification textures and 94 secondary minerals. Lithic particles are characterized by their dark, dull appearance, and non-vesicular 95 density with incipient alteration and rounded edges. The classification of some of the particles 96 sometimes included Scanning Electron Microscope (SEM) analysis to examine for etch pits or 97 alteration signs. Juvenile particles are distinguished by their gloss, smooth-skinned surface and lack of 98 alteration. SEM analyses are sometimes necessary to ascertain the lack of alteration signs ^{6,15,16}.

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100 Database design. The database is structured using MongoDB, a NoSQL database that organizes data 101 into flexible, schema-less collections. These collections store and manage various types of volcanic 102 data (Table 1). They include the Volcanoes and Eruptions collections (Figure 2) which record basic 103 information such as the vent location, eruption date, main rock compositions, and are sourced from Smithsonian's "Volcanoes of the World" database¹⁷. The Ash-Forming Events (AFEs) collection records 104 105 information about the ash-forming event such as the date and its volcanic context (e.g., eruptive style). 106 The Samples collection contains information about sampling details such as the location and 107 technique, and lab procedures such as cleaning and sieving. The Particles collection contains a unique 108 identifier for each particle with its associated image, metadata such as the imaging conditions, and 109 the 33 measured *features*.

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111 Architecture and Technologies of VolcAshDB web-based platform. The platform consists of two main 112 parts, a data input of images from which the features are extracted, and a data output that allows for 113 search, visualization, analysis and downloading of the data (Figure 3). The platform is built using the 114 MERN stack, which includes MongoDB, Express.js, React.js, and Node.js. The backend uses Node.js 115 and Express.js for server-side functions, while MongoDB stores and manages the data. The frontend 116 is built with React.js, providing a responsive interface for easy interaction and visualization. 117 Communication between the client and server is handled through a RESTful API. The platform runs on 118 a Linux server managed by Proxmox VE, with deployment managed by PM2 for process control.

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120 Data Records

Access. VolcAshDB contents are publicly accessible for image and feature visualization and plotting through the VolcAshDB interface at https://volcashdb.ipgp.fr/catalogue, and for download after

- logging in. To sign up, users must provide an email for verification. The data are licensed under CC-BY4.0 and can be cited using the DOI <u>https://doi.org/10.18715/ipgp.2024.lx32oxw9</u>.
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126 Data coverage. As of September 2024, VolcAshDB contains the 33 features and classified image data 127 from 12,044 volcanic ash particles from 53 samples across 13 different volcanoes (Table 2). The 128 collection comprises basaltic to rhyolitic magmas, and a range of activity types, including phreatic 129 eruptions, lava dome explosions, basaltic lava fountaining, and Plinian to Sub-plinian eruptions. The 130 database also includes experimental samples that provide insights into the effect of alteration of the particles under a range of specific conditions of temperature and oxygen fugacity¹⁸. Some of our 131 samples have also been used to better understand volcanic processes during activity, such as for 132 Nevados de Chillán eruptive period, 2016-2018¹⁹, or are currently being investigated to obtain critical 133 insights into the eruption progression at Marapi, 2023–2024 (Nurfiani et al., in prep). 134

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136 **Technical Validation**

137 Effect of image resolution and focus on measured features. To evaluate how the features extracted 138 from the particle images are affected by poor resolution or focus, we modeled how the feature values 139 change by varying the two parameters on our particle images. For this we used four particle images 140 with different shape, texture and color. The resolution unit we use to model is defined by the number 141 of pixels (pxls) comprised in the area (a) of the particle per image (pxls/a). We found that feature 142 values only begin to shift when resolution falls below 3.5x104 pxls/a, whereas our average image 143 resolution is 1.41x106 pxls/a, confirming that our resolution is more than sufficient (Figure 4A). To 144 evaluate the effect of focus, we applied a Gaussian blur to the images (measured in sigmas). We found 145 that feature values remain stable until the blur reaches 0.2 sigma, beyond which convexity starts to change and some instances become truncated (Figure 4B). This shows that extracted features in our 146 147 database are not influenced by poor focus.

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149 Usage Notes

150 Catalogue, Image Browsing and Data Download. Users can access the page "Catalogue" at 151 https://volcashdb.ipgp.fr/catalogue to browse images. The catalogue contains two subsets of data, 152 those of natural samples and those of experimental samples. Users can swap between them by ticking 153 the "Natural Data" button. Advanced image searching can be done by using a set of filters displayed 154 as tags with dropdown menus (Figure 5). Filters include options for volcano, eruptive activity type, 155 grain-size fraction, particle type and shape. To download the filtered images, particle features and 156 metadata, users must first sign up by providing an email address, country, and institute. Upon email 157 verification, users will be able to click on "Apply Filters" and "Download Images and Measured Features". A zip file is automatically created which can be downloaded with the filtered particle images 158 159 and an excel spreadsheet. The spreadsheet contains the particle features and metadata (imaging 160 conditions, volcano, date of emission, etc.) and each row represents a particle of the dataset.

161 162 Data analysis and Plots. Users can access https://volcashdb.ipgp.fr/analytic to visualize and interact with various plots that illustrate the contents of the database in various manners (Figure 6). Plots are 163 164 provided for natural or experimental samples, and users can swap between them by ticking the 165 "Natural Data" button. Pie charts show the proportions of number of particles per volcano, per particle 166 type and per activity type, and users can select or unselect specific subgroups. Juvenile-Altered 167 material-Lithic ternary diagrams show the proportion of particle types normalized without free crystals per sample and according to eruptive activity type. We also analyze by Principal Component 168 Analysis the features as reported in Benet et al.⁹ and show the data transformed into the Principal 169 Components 1 and 2. The plot is accompanied with a dropdown menu that allows users to select 170 171 among eruptive activity types. The "Subplinian", for instance, shows a distinctive cluster of juvenile 172 particles in violet color. Histograms show the distribution of values of a given feature and two

- dropdown menus allow choosing which feature and by which volcano. Further insights from all plotscan be obtained by hovering with the mouse over any marker.
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176 Code Availability

The Python program for feature extraction and for Principal Component Analysis are available at <u>https://github.com/dbenet-ntu/Volcash-Project</u>. The code for the web-based platform is available at <u>https://github.com/dbenet-ntu/VolcAshDB-web_Sept2024</u>. The Python code was developed using its version 3.9. Data wrangling and analysis was done using Numpy v.1.21.6, Pandas v. 1.3.5, image processing using Scikit-image v.0.19.2 and opencv v.3.4.2, and plotting by matplotlib v.3.5.1 and seaborn v.0.11.2.

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189190 Author contributions

191 Benet led the development, acquired and processed the data, and wrote the original manuscript of 192 the project, which was supervised and revised by Costa. Migadel and Benet led the development of

193 front-end and the back-end of the web-platform. Lee, D'Oriano, Pompilio, Nurfiani and Rifai provided

194 some samples. The final version of the manuscript has been read, reviewed and approved by all co-195 authors.

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197 **Competing interests**

198 The authors declare no competing interests.

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Figure 2







218 Figure Legends

219 Figure 1: Summary of the main contents and data types found in VolcAshDB. (A) World map²⁰ with the 220 location of volcanoes for which we have collected data, and pie charts showing the average proportion 221 of the particle main types (orange color is altered material, blue is free-crystal, violet is juvenile and 222 green is lithic) of samples by volcano. (B) Examples of the properties we have measured for each ash 223 particle, including the contour for shape, the grayscale homogeneity of the surface for texture, and 224 the color-channels histogram distributions for color. (C) Plots showing the varied distributions of three 225 particle features (convexity, correlation, mode of saturation) across the four main types (color-coded 226 as above). "Density", in the y-axis, is the probability density function calculated by the Kernel Density 227 Estimate plot, which can take values above 1.

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Figure 2. Schematic of the database hierarchical design based on five MongoDB *collections*. Lists within bullet points are not exhaustive. *Volcanoes* and *Eruptions collections* were adopted from Smithsonian's "Volcanoes of the World" database (Global Volcanism Program, 2024).

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- Figure 3. Dataflow, main tasks and general architecture of the web-based platform.

Figure 4. Plots with the modelled feature values to estimate the influence of image focus and resolution. (A) The values of the features standard deviation of saturation, contrast, solidity are modelled by decreasing pixel resolution, and (B) the values of saturation mean, homogeneity, convexity are modelled at varying blur for four representative particle images with grain-size between 1–2 mm. Colors of the curves in (A) and (B) correspond to the different particles shown inside the

- "Contrast" panel. The effect of the Gaussian blur to the original image (top-left image) is shown at 5
 (top-center) and 10 (top-right) sigmas. Dashed lines indicate the values of resolution or focus beyond
 which feature values start to shift.
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- Figure 5. Screenshot of the "Catalogue" page in VolcAshDB web site. The screenshot includes the following filters: "Volcano Name" as "La Palma", "Main Type" as "juvenile", "Shape" as "fluidal" and "Crystallinity" as "low". The white bars show the scale for 1 mm under the microscope.
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- 249 Figure 6. Screenshots of six representative plots from the "Plots" page in VolcAshDB web site.
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251 **Tables**

Table 1: List of fields and definitions of the collections that make VolcAshDB.

Field	Туре	Optionality	Description			
Volcanoes collection (source Smithsonian database)						
_id	ObjectId	O ¹	Document identifier			
volc_name	String	M ²	Volcano name			
volc_num	Int(32)	Μ	Volcano identifier			
volc_slat	String	М	Volcano location (latitude)			
volc_slon	String	Μ	Volcano location (longitude)			
volc_status	String	0	Status of the volcano			
volc_type	String	0	Type of volcano			
Eruptions collection	(source Smith	sonian datab	ase)			
_id	ObjectId	0	Document identifier			
volc_num	Int(32)	Μ	Volcano identifier			
volc_name	String	Μ	Volcano name			
ed_stime	Date	Μ	Eruption start date			
ed_etime	Date	Μ	Eruption end date			
Ash-Forming Events Co	ollection					
_id	ObjectId	0	Document identifier			
afe_code	String	М	Ash forming event identifier			
volc_num	Int(32)	М	Volcano identifier			
afe_date	Date	Μ	Ash forming event date			
eruptive_style	String	Μ	Description of eruption style			
afe_lat	String	Μ	Ash forming event location (latitude)			
afe_lon	String	М	Ash forming event location (longitude)			
plume_col	String	0	Plume color			
max_plume_height	Int(32)	0	Maximum recorded plume height			
fumarolic_activity	String	0	Observed fumarolic activity intensity			
Samples collection						
_id	ObjectId	0	Document identifier			
sample_stime	Date	Μ	Sampling date			
sample_lat	String	Μ	Sampling coordinate latitude			
sample_lon	String	Μ	Sampling coordinate longitude			

sample_nat	Boolean	М	Sample is natural or experimental
sample_techn	String	0	Technique (e.g. ashmeter)
sample_surf	String	0	Sampling surface (e.g. roof)
sample_collector	String	0	Collector/s name
ultrasound	Boolean	0	Ultrasonicated
gsLow	String	Μ	Lower value of grain-size fraction
gsUp	String	Μ	Upper value of grain-size fraction
Particles collection			
_id	ObjectId	0	Document identifier
imgURL	String	Μ	Filename of the particle's image
main_type	Object	Μ	Main type of the particle
sub_type	String	0	Sub type of the particle
color	String	0	Color of the particle
luster	String	0	Luster of the particle
shape	String	0	Shape of the particle
crystallinity	String	0	Crystallinity of the particle
alter_degree	String	0	Degree of alteration of the particle
weathering_sign	String	0	Presence of weathering signs
instrument	String	0	Instrument for particle image
magnification	Int(32)	0	Magnification of particle image
multi_focus	Boolean	0	Use of multi-focus for particle image
convexity ³	Double	М	Convexity of the particle (shape)
contrast ³	Double	М	Contrast of the particle (texture)
hue_mean ³	Double	М	Mean value of hue histogram (color)

253 1: "O" stands for "Optional"

254 2: "M" stands for "Mandatory"

255 3: Only one feature for each category of shape, texture and color is listed as example. See Benet et al

⁹ for more details.

257

258 Table 2: Characteristics of the volcanoes represented in our database and number of particles that

259 have been analyzed and archived.

Volcano	Magma composition	Volcano type	Reference	# particle images
Cumbre Vieja	Trachybasalt / Tephrite Basanite	Cinder cone	21	1075
Kelud	Andesite / Basaltic Andesite	Stratovolcano	22	665
Merapi	Basaltic Andesite	Stratovolcano	23	1098
Soufrière de Guadeloupe	Basaltic andesite- Andesite	Stratovolcano	24	484
Nevados de Chillán	Basaltic-Andesite	Dome complex	25	1438
Ontake	Andesite	Stratovolcano	26	777
Pinatubo	Dacite	Caldera	27	506

Mount St. Helens	Dacite	Stratovolcano	28	261
Marapi	Andesite	Stratovolcano	29	2776
Toba	Dacite / Rhyolite	Caldera	30	40
Vesuvius (EXP) ¹	Tephriphonolite	Stratovolcano	18	806
Etna (EXP)	Trachybasalt / Tephrite Basanite	Stratovolcano	18	1374
Stromboli (EXP)	Basalt	Stratovolcano	18	645

260 1: (EXP) denotes that the sample has been experimentally obtained for example by crushing scoria into ash-sized grains ¹⁸. 261 262 References 263 264 Self, S. Effects of volcanic eruptions on the atmosphere and climate. in (eds. Joan Martí & 1. 265 Gerald G. J. Ernst) (Cambridge: Cambridge University Press, 2005). 266 2. Bonadonna, C., Biass, S., Menoni, S. & Gregg, C. E. Assessment of risk associated with tephra-267 related hazards. in Forecasting and Planning for Volcanic Hazards, Risks, and Disasters 329-268 378 (Elsevier, 2020). doi:10.1016/B978-0-12-818082-2.00008-1. 269 3. Horwell, C. J. et al. A physico-chemical assessment of the health hazard of Mt. Vesuvius volcanic 270 ash. Journal of Volcanology and Geothermal Research 191, 222–232 (2010). 271 4. Miwa, T., Geshi, N. & Shinohara, H. Temporal variation in volcanic ash texture during a 272 vulcanian eruption at the sakurajima volcano, Japan. Journal of Volcanology and Geothermal 273 Research 260, 80-89 (2013). 274 5. Suzuki, Y. et al. Precursory activity and evolution of the 2011 eruption of Shinmoe-dake in 275 Kirishima volcano-insights from ash samples. *Earth, Planets and Space* **65**, 591–607 (2013). 276 6. Gaunt, H. E. et al. Juvenile magma recognition and eruptive dynamics inferred from the analysis 277 of ash time series: The 2015 reawakening of Cotopaxi volcano. Journal of Volcanology and 278 Geothermal Research 328, 134–146 (2016). 279 7. Cashman, K. V. & Hoblitt, R. P. Magmatic precursors to the 18 May 1980 eruption of Mount St. Helens, USA. Geology 32, 141-144 (2004). 280 281 8. Pardo, N. et al. Perils in distinguishing phreatic from phreatomagmatic ash; insights into the 282 eruption mechanisms of the 6 August 2012 Mt. Tongariro eruption, New Zealand. Journal of 283 Volcanology and Geothermal Research 286, 397–414 (2014). 284 9. Benet, D. et al. VolcAshDB: a Volcanic Ash DataBase of classified particle images and features. 285 Bull Volcanol 86, (2024). 286 10. Benet, D., Costa, F. & Widiwijayanti, C. Volcanic Ash Classification Through Machine Learning. 287 Geochemistry, Geophysics, Geosystems 25, (2024). 288 11. Liu, E. J., Cashman, K. V. & Rust, A. C. Optimising shape analysis to quantify volcanic ash 289 morphology. GeoResJ 8, 14-30 (2015). 290 12. Dürig, T. et al. Particle shape analyzer Partisan - An open source tool for multi-standard two-291 dimensional particle morphometry analysis. Annals of Geophysics 61, (2018). 292 Nurfiani, D. & Bouvet de Maisonneuve, C. Furthering the investigation of eruption styles 13. 293 through quantitative shape analyses of volcanic ash particles. Journal of Volcanology and 294 Geothermal Research **354**, 102–114 (2018). 295 14. Haralick, R. M., Dinstein, I. & Shanmugam, K. Textural Features for Image Classification. IEEE 296 *Trans Syst Man Cybern* **SMC-3**, 610–621 (1973). 297 15. D'Oriano, C. et al. Syn-Eruptive Processes During the January-February 2019 Ash-Rich 298 Emissions Cycle at Mt. Etna (Italy): Implications for Petrological Monitoring of Volcanic Ash. 299 Front Earth Sci (Lausanne) 10, (2022).

- 30016.D'Oriano, C., Bertagnini, A., Cioni, R. & Pompilio, M. Identifying recycled ash in basaltic301eruptions. Sci Rep 4, (2014).
- 302 17. Global Volcanism Program, 2024. [Database] Volcanoes of the World (v. 5.2.2; 22 Aug 2024).
 303 Distributed by Smithsonian Institution, compiled by Venzke, E.
 304 https://doi.org/10.5479/si.GVP.VOTW5-2024.5.2. (2024).
- 18. D'Oriano, C., Pompilio, M., Bertagnini, A., Cioni, R. & Pichavant, M. Effects of experimental
 reheating of natural basaltic ash at different temperatures and redox conditions. *Contributions* to Mineralogy and Petrology 165, 863–883 (2013).
- Benet, D., Costa, F., Pedreros, G. & Cardona, C. The volcanic ash record of shallow magma intrusion and dome emplacement at Nevados de Chillán Volcanic complex, Chile. *Journal of Volcanology and Geothermal Research* **417**, (2021).
- 31120.Natural Earth. 2023. 'Admin 0 Countries.' 1:10m scale. Natural Earth. Accessed August 19,3122024. https://www.naturalearthdata.com/downloads/10m-cultural-vectors/10m-admin-0-313countries/. earth_map_ref. (2023).
- Romero, J. E. *et al.* The initial phase of the 2021 Cumbre Vieja ridge eruption (Canary Islands):
 Products and dynamics controlling edifice growth and collapse. **431**, (2022).
- Utami, S. B., Costa, F., Lesage, P. H., Allard, P. & Humaida, H. Fluid fluxing and accumulation
 drive decadal and short-lived explosive basaltic andesite eruptions preceded by limited
 volcanic unrest. *Journal of Petrology* 62, 1–29 (2021).
- Gertisser, R. *et al.* Geological History, Chronology and Magmatic Evolution of Merapi. in *Active Volcanoes of the World* 137–193 (Springer Science and Business Media Deutschland GmbH,
 2023). doi:10.1007/978-3-031-15040-1_6.
- 322 24. Metcalfe, A. *et al.* Magmatic Processes at La Soufrière de Guadeloupe: Insights From Crystal
 323 Studies and Diffusion Timescales for Eruption Onset. *Front Earth Sci (Lausanne)* 9, (2021).
- 324 25. Dixon, H. J. *et al.* The geology of Nevados de Chillan volcano, Chile. *Revista Geologica de Chile* 325 26, 227–253 (1999).
- 32626.Miyagi, I., Geshi, N., Hamasaki, S., Oikawa, T. & Tomiya, A. Heat source of the 2014327phreatic eruption of Mount Ontake, Japan. Bull Volcanol 82, (2020).
- 32827.Bernard, A. et al. Petrology and Geochemistry of the 1991 Eruption Products of Mount329Pinatubo. Fire and mud; eruptions and lahars of Mount Pinatubo, Philippines (1996).
- Scheidegger, K. F., Federman, A. N. & Tallman, A. M. Compositional heterogeneity of tephras
 from the 1980 eruptions of Mount St. Helens. *J Geophys Res Solid Earth* 87, 10861–10881
 (1982).
- Nurfiani, D. *et al.* Combining Petrology and Seismology to Unravel the Plumbing System of a
 Typical Arc Volcano: An Example From Marapi, West Sumatra, Indonesia. *Geochemistry, Geophysics, Geosystems* 22, (2021).
- 336 30. Chesner, C. A. & Rose, W. I. Vol a Olo Stratigraphy of the Toba Tuffs and the Evolution of the
 337 Toba Caldera Complex, Sumatra, Indonesia. vol. 53 (1991).
- 338