Probabilistic Volcanic Hazard Assessment of the 22.5–28°S Segment of the Central Volcanic Zone of the Andes

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Evaluation of volcanic hazards typically focusses on single eruptive centres or spatially restricted areas, such as volcanic fields. Expanding hazard assessments across wide regions (e.g., large sections of a continental margin) has rarely been attempted, due to the complexity of integrating temporal and spatial variability in tectonic and magmatic processes. In this study, we investigate new approaches to quantify the hazards of such long-term active and complex settings, using the example of the 22.5–28°S segment of the Central Volcanic Zone of the Andes. This research is based on the estimation of: 1) spatial probability of future volcanic activity (based on kernel density estimation using a new volcanic geospatial database), 2) temporal probability of future volcanic events, and 3) areas susceptible to volcanic flow and fall processes (based on computer modeling). Integrating these results, we produce a set of volcanic hazard maps. We then calculate the relative probabilities of population centres in the area being affected by any volcanic phenomenon. Touristic towns such as La Poma (Argentina), Toconao (Chile), Antofagasta de la Sierra (Argentina), Socaire (Chile), and Talabre (Chile) are exposed to the highest relative volcanic hazard. In addition, through this work we delineate five regions of high spatial probability (i.e., volcanic clusters), three of which correlate well with geophysical evidence of mid-crustal partial melt bodies. Many of the eruptive centres within these volcanic clusters have poorly known eruption histories and are recommended to be targeted for future work. We hope this contribution will be a useful approach to encourage probabilistic volcanic hazard assessments for other arc segments.

Keywords: Central Volcanic Zone, Chile, Argentina, MatHaz, volcanic hazards
INTRODUCTION

Traditionally, volcanic hazards are assessed at the scale of single volcanic systems, such as stratovolcanoes (e.g., Ogburn et al., 2020), or single volcanic fields (e.g., Connor and Hill, 1995). Motivation for hazard assessments typically stems from a perception of future risk, based on past eruptive casualties, rapid population and infrastructure growth, insights from new age or other geological information, and/or an increase in unrest indicators (e.g., Siebe and Macías, 2003; Bird et al., 2010; Sandri et al., 2014; Biass et al., 2016; Charbonnier et al., 2020; Jiménez et al., 2020). Volcanic hazard assessments depend upon assembling detailed information on eruption frequency, magnitude and style, integrating information from diverse geomorphic, structural, geochronological, geophysical, and geochemical data (e.g., Clavero et al., 2004; Harpel et al., 2011; Johnston et al., 2011; Alloway et al., 2017).

Different tools are required for assessing volcanic hazards at distributed volcanic fields (e.g., Bartolini et al., 2015; Sieron et al., 2021), volcanic islands (e.g., Becerril et al., 2014; Cappello et al., 2016), caldera systems (e.g., Bevilacqua et al., 2015; Charlton et al., 2020), and large complex multi-vent volcanoes (e.g., Cappello et al., 2012; Jiménez et al., 2018). The greatest problem in dealing with such complex volcanic systems is that the distribution of specific eruptive hazards needs to be examined in relation to the locations of possible vent opening (e.g., Sandri et al., 2018; Bevilacqua et al., 2021). In addition to investigating spatial variability, long-lived volcanism at complex sites generates strong temporal hazard variability, which is a challenge to understand in contemporary forecasts (e.g., Bebbington and Cronin, 2011; Lerner et al., 2019).

The problems of integrating spatial and temporal variations in hazard are compounded when examining larger areas of interest, such as volcanic arcs. Past regional-scale assessments typically analysed every volcano independently (e.g., Mullineaux, 1976; Miller, 1989; Lara et al., 2011; Amigo et al., 2012). However, this is only feasible when a handful of volcanoes in a region are considered. If dozens or hundreds of volcanic centres are analysed, new approaches and methods are needed. Techniques have been developed to forecast the location and timing of future volcanic vent opening in arc segments (Martin et al., 2004; Jaquet et al., 2012; Jaquet et al., 2017; Germa et al., 2013; Kósik et al., 2020; O’Hara et al., 2020), which is the first step towards regional assessments. There is a further need to integrate spatial and temporal histories into these types of analyses, along with a comprehensive and systematic methodology to evaluate different types, frequencies and scales of volcanic processes associated with each vent/volcano. Another challenge when carrying out volcanic hazard assessments in large and complex volcanic regions is the inherent difficulty of collecting comprehensive data that are of similar quality and resolution across a broad area. Even data coverage, such as in age information, mapping or volume estimations are less likely (and less affordable) as the spatial scope increases to whole arcs or long-lived continental margin segments.

Despite the complex challenge that the assessment represents, evaluating volcanic hazards over very large areas can be useful for a first-order synoptic assessment to enable the targeting of limited resources for new investigations or mitigation efforts to the most important areas. Furthermore, a regional long-term analysis can help to understand larger system changes that may highlight critical features for future hazard, such as regional migration of volcanic focussing, arc-scale variations in volcanic flux, and other specific regional-based tectonic or other drivers for volcanic flare-ups.

In consequence, in this study we examine approaches to understand the relative spatial and temporal aspects of volcanic hazards within a large continental margin setting. This study is based on an area between latitudes 22.5–28°S along the Central Volcanic Zone of the Andes of South America (Figure 1). This region records ~35 Myr of uninterrupted volcanic activity, comprising a vast variety of activity types, volcanic landforms and erupted compositions. It includes the site of one of the largest (VEI 6+) Holocene explosive eruptions in the Andes, which occurred approximately 4.2 ka BP from the Cerro Blanco volcanic complex (26.8°S/67.7°W) (Báez et al., 2015; Fernández-Turiel et al., 2019; Báez et al., 2020a; de Silva et al., 2022), whereas the only confirmed historical (i.e., post ~1750 AD) eruptions in this arc segment occurred at Láscar (23.4°S/67.7°W), with the largest event (VEI of 3–4) in April 1993 (Gardeweg and Medina, 1994). However, much of the volcanic history of this area remains poorly constrained, even though many morphologically young volcanic features can be recognized throughout the region and several of these are considered potentially active (de Silva and Francis, 1991; González-Ferrán, 1995).

Here, we merge the MatHaz tool of Bertin et al. (2019) with a new volcanic geospatial database (Bertin et al., 2022) to generate the first probabilistic volcanic hazard assessment for the ~22.5–28°S segment of the Central Volcanic Zone of the Andes. MatHaz is an open-source computer code that performs an integrated probabilistic volcanic hazard assessment in regions characterized by distributed volcanism, whereas the Bertin et al. (2022) database collates, for the study area, all available late Cenozoic (<36 Ma) information on volcano types, vent locations, eruptive volumes and eruption ages. By integrating these two resources, we identify the most likely regions to host future volcanic activity, and delineate areas likely to be affected by pyroclastic density currents, ballistic projectiles, lava flows, debris flows, and tephra fallout. In doing so we are able to recognize, for the first time, order-of-magnitude variations in volcanic hazards across the region. After integrating these results into a single hazard map, we evaluate the relative volcanic hazard for all the settlements of the region. Overall, we hope the approach and results from our study will provide a guide for more targeted hazard and risk assessments in the future.

TECTONIC SETTING OF THE ~22.5–28°S SEGMENT

The ~22.5–28°S segment of the western margin of South America is part of the Central Volcanic Zone (CVZ) of the Central Andes...
The Central Andes is an orogenic belt located between ~5 and ~33°S that includes the CVZ and two flat slab segments (Peruvian to the north and Pampean to the south). The CVZ spans between ~13 and ~28°S, and includes active and potentially active volcanoes of southern Peru, northern Chile, western Bolivia and northwestern Argentina. Volcanism in this region can be traced as far back as the Early Jurassic (190–180 Ma) (e.g., Parada et al., 2007; Ramos, 2009). Volcanic activity was semi-continuous since then, and its front migrated inland from the Coastal Cordillera (Early Jurassic-Late Cretaceous arc) to the Precordillera (Late Cretaceous-Paleogene arc), and then to the Western Cordillera (Neogene-Quaternary arc) (e.g., James and Sacks, 1999; Haschke et al., 2002; Charrier et al., 2007; Hoke and Lamb, 2007) (Figure 1B). During the middle Miocene, volcanic activity broadened to the east across much of the Altiplano-Puna (a hinterland plateau), and then progressively retreated back to the Western Cordillera during the upper Miocene and Pliocene (e.g., Coira et al., 1993; Allmendinger et al., 1997; Trumbull et al., 2006; Kay and Coira, 2009; Guzmán et al., 2014).

The magmatic broadening and narrowing event has been linked to a transient shallowing of the subduction zone due to the passage of the Juan Fernández Ridge (e.g., Trumbull et al., 2006; Kay and Coira, 2009; Freymuth et al., 2015; de Silva and Kay, 2018). This ridge, along with the Taltal and Copiapó oceanic ridges, earlier caused a flat slab event that affected a large segment of the Central Andean margin centred at ~18–20°S during the Eocene-Oligocene (45–25 Ma) (James and Sacks, 1999; O’Driscoll et al., 2012; Bello-González et al., 2018; Martinod et al., 2020). The passage of the Juan Fernández Ridge was followed 5–10 Myr later by a period of large-volume volcanism (e.g., Kay and Coira, 2009; Kay et al., 2010; Beck et al., 2015; Freymuth et al., 2015; Brandmeier and Wörner, 2016), which resulted in the Central Andean Neogene Ignimbrite Province (Kay et al., 2010; de Silva and Kay, 2018). In particular, the Altiplano-Puna Volcanic Complex (APVC; de Silva, 1989) and Cerro Galán (Kay et al., 2011) ignimbrite provinces are prominent foci in the study area (Figure 1B). The first-order correlation between the passage of the Juan Fernández Ridge and large-volume volcanism has given rise to the proposal that the southernmost segment of the CVZ (i.e., around ~27–28°S) has the potential to host a magmatic flare-up in the future (e.g., Kay and Coira, 2009; Beck et al., 2015; Ward et al., 2017). In addition, along with the arc broadening and narrowing, back-arc magmatism also occurred in this region in response to lithospheric foundering events (e.g., Kay et al., 1994; Risse et al., 2013; Mulcahy et al., 2014; Maro et al., 2017; Chen et al., 2020; Gao et al., 2021). Collectively, volcanism has been controlled by a range of varying tectonic processes (e.g., Tibaldi et al., 2017; Naranjo et al., 2018; Morfulis et al., 2020).

Currently, Láscar is the most active volcano in this segment of the CVZ (Figure 2), with 34 eruptions recorded between 1848...
and October 2015 (Casertano, 1963; González-Ferrán, 1995; Petit-Breuilh, 2004; Global Volcanism Program, 2016). Láscar has also produced the only VEI 3-4 eruption in this arc segment in the historical record (i.e., since ~1750 AD), in April 1993. Other volcanoes with unconfirmed historical eruptions include (Figure 2): Llullaillaco, with seven events between 1854 and 1960 (Casertano, 1963; Petit-Breuilh, 2004); Aracar, with one event in March 1993 (Global Volcanism Program, 1993a); and Ojos del Salado, with one event in November 1993 (Global Volcanism Program, 1993b). Four volcanoes host long-term active fumarolic activity: Alítar, Láscar, Lastarria, and Ojos del Salado (de Silva and Francis, 1991; González-Ferrán, 1995), four are actively deforming: Lastarria, Cordón del Azufre, Cerro Bayo, and Cerro Blanco (Pritchard and Simons, 2002; Henderson and Pritchard, 2013), and nine have shown thermal anomalies detected by infrared satellite imagery: Licancabur, Alítar, Láscar, Chiquiques, Púlar-Pajonales, Lastarria, Sierra Nevada, Falso Azufre, and Ojos del Salado (Jay et al., 2013). Only two of these volcanoes, Láscar and Lastarria, are currently actively monitored (Amigo, 2021; Aguilera et al., 2022).

Eruption histories of only a few volcanoes in the region have been comprehensively studied (e.g., Richards and Villeneuve, 2001; Richards et al., 2006; Naranjo, 2010; Gardeweg et al., 2011; Norini et al., 2014; Grosse et al., 2018, Grosse et al., 2022; Bustos et al., 2019); overall, they reveal a rich history of activity ranging in type (from effusive to explosive), volume (from \(1 \times 10^{-3}\) to 10 s of km\(^3\) of magma), and composition (from basaltic andesites to rhyolites). Specific vents at Láscar, Lastarria, and Cerro Blanco have been thoroughly studied and are known to pose high hazard (Perucca and Moreiras, 2009; Amigo et al., 2012;
Gardeweg and Amigo, 2015; Elissondo et al., 2016). However, these vents are a small fraction of the hundreds of youthful-looking existing vents in the area. The lack of detailed information on these other vents has hindered systematic regional hazard assessment. Recent studies have just begun to assess specific hazards in this area, including: pyroclastic density currents (Cerro Blanco; Báez et al., 2020a; Báez et al., 2020b), ballistic projectiles (Láscar; Bertin, 2017), lava flows (Chorrillos and San Gerónimo; Fernández-Turiel et al., 2021), debris flows (Lastarria; Rodríguez et al., 2020), and tephra fallout (Cerro Blanco; Fernández-Turiel et al., 2019). Building on this, a regional perspective on volcanic hazards is needed in order to identify which volcanic regions are most likely to become active in the future, which areas will be affected the most by future volcanic activity, and to motivate additional studies to better understand hazards at more detailed scales. The regional hazard assessment presented here is based on a comprehensive volcanic database and simplified tools for modeling volcanic hazards.

**METHODS**

Volcanic hazard assessments for specific segments of the CVZ (Perucca and Moreiras, 2009; Amigo et al., 2012; Ayala, 2014; Macedo et al., 2016) have typically analysed each volcano independently and conditional upon an eruption occurring. However, this approach is best suited for comparison of a handful of volcanoes. If hundreds of volcanic centres are analysed, as it is the case here, different approaches and methods are needed to take spatial and temporal variations in hazard into account. On the broadest scale, different techniques have been developed to forecast the location and timing of future volcanic activity in arc segments (e.g., Martin et al., 2004; Jaquet et al., 2012; Germa et al., 2013; Jaquet et al., 2017; Kósik et al., 2020; O’Hara et al., 2020). Of these, the approach of Martin et al. (2004) is particularly useful for the purposes of our application as it performs an estimation of the spatio-temporal hazard by modeling each volcanic vent as monogenetic or polygenetic. Hence, following the Martin et al. (2004) approach, we classified the volcanic vents within our study area (vents extracted from the Bertin et al., 2022 database) according to their categories (Németh and Kereszturi, 2015): monogenetic (single event in unique site), transitional-polycyclic (multiple events but each with a unique magma driver), or polygenetic (multiple events driven by a long-lived magmatic feeding system), based on the amount and type of different volcanic phenomena sourced from each vent (see [Supplementary Table S1](#) for the criteria followed). We concentrated our analysis on the last 36 Myr, postdating the latest major magmatic migration of the arc.

The Bertin et al. (2022) database lists 2,950 volcanic vents and 2,057 volcanic deposits (e.g., ignimbrite sheets, lava flows). For simplicity, to estimate vents for those volcanic deposits without an evident source location, 479 additional vents were mapped at the geometric centres of the deposits; thus, 3,429 vents were identified in total (1,825 monogenetic, 133 transitional-polycyclic, 1,471 polygenetic). For every volcanic vent, maximum and minimum ages (expressed as a mean and two standard deviations) were assigned based on the radiometric ages of its associated volcanic deposits. For volcanic deposits lacking radiometric ages, estimations were conducted based on the ages of the adjoining deposits with a clear stratigraphic relationship. If this was not possible, a broad estimate was made based on the ages of all neighbouring deposits with similar geomorphic preservation. The volcanic vents layer thus created is available in geospatial vector data format in [Supplementary Material S1](#).

From Vents to Events

A major challenge in volcano science is to unravel the eruptive histories of volcanoes from the geologic record (e.g., Damaschke et al., 2018; Lerner et al., 2019; Pure et al., 2020). On an arc-scale, we cannot hope, at least at this point in time, to successfully integrate the volcanic histories of hundreds of volcanoes. Therefore, here we estimate the average number of volcanic events per vent type over a fixed time interval assuming a fixed erupted volume.

An additional hindrance is that, even for ‘monogenetic’ vents, it is not certain that each volcanic vent represents a single eruptive episode (Runge et al., 2014; Báez et al., 2016; Bevilacqua et al., 2017; Connor et al., 2018; Gallant et al., 2021). This uncertainty, along with the likely presence of buried and eroded vents, can be examined through statistical (Runge et al., 2014), morphometric (Nieto-Torres and Martin Del Pozzo, 2019), and spatio-temporal clustering (Gallant et al., 2021) techniques. However, due to the large study area and the availability of age and volume information in the Bertin et al. (2022) database (each volcanic deposit includes an age and volume range), here we followed an alternative approach. For each volcanic deposit within the database (excluding voluminous ignimbrites), a specific volume was estimated from the known volume ranges assuming a uniform distribution. These volumes were summed to obtain a total erupted volume. Using a Monte Carlo approach, the volume estimations were repeated $10^6$ times, constraining the total volume of the arc segment to be between 15,800 and 16,400 km$^3$. This volume range was then used to forecast the total number of volcanic events assuming the following: 1) age ranges were calculated for each vent by sampling their maximum and minimum ages; 2) every volcanic event had a volume of 0.1 km$^3$, which is the total dense rock equivalent (DRE) volume calculated for the Láscar 1993 eruption (Calder et al., 2000; Gardeweg et al., 2011); and 3) a fixed time interval of 10 kyr from 36 Ma to present (i.e., 3,600 time bins), deemed here as a sensible partition to evaluate event recurrence rates in detail without dramatically losing computational efficiency. Monogenetic vents were assumed to have only one eruption, polycyclic vents were assumed to have one eruption every 10 kyr, and polygenetic eruption rates were unknown. Hence, we estimated the polygenetic eruption rate by trial and error. That is, we started with an arbitrary polygenetic eruption rate and calculated the total number of events $10^4$ times. When the mean of the total number of events, multiplied by 0.1 km$^3$, was within the range 15,800–16,400 km$^3$, then we kept that polygenetic eruption rate and repeated the exercise with a slightly different eruption rate until the whole volume range was covered. The code
used to perform these calculations is available in Supplementary Material S2.

The results indicate a rate of 0.4433 (−0.0078/+0.0081) events per 10 kyr per polygenetic vent. This means that for all post-10 ka potentially active vents (137), an estimated 45.77 (−1.41/+1.35) events occurred, i.e., a 218 (−6/+7) year interval between 0.1 km³ DRE events, a reasonable result taking into account that since AD ~1750, only one 0.1 km³ eruption has been recorded in the region. The values obtained above, however, change when other event volumes are considered. This issue, and how it can influence the hazard analysis, is further addressed in the discussion.

Topographic Data
In order to provide a topographic digital base onto which conduct the hazard simulations, we compiled 135 ASTER GDEM v3 files (Advanced Spaceborne Thermal Emission and Reflection Digital Elevation Model; https://search.earthdata.nasa.gov). Each square degree tile has an average spatial resolution of ~30 m and a vertical accuracy of 7–14 m (Talchabahdel et al., 2021). Using Global Mapper v18 (https://www.bluemarblegeo.com/global-mapper/), we clipped out the region inside 22–30°S/70°W and reduced the spatial resolution to 1,000 m in order to generate a manageable file (<10 Mb). This pixel size was used to adjust the coordinates of each vent to make them compatible with the grid. Finally, the topography was exported as an Excel file (Supplementary Material S3) to be loaded in MatHaz.

Spatial and Temporal Volcanic Hazard Assessments
MatHaz provides a semi-automated probabilistic volcanic hazard assessment in regions characterized by distributed volcanism (Supplementary Material S7). The spatial analysis was conducted using an elliptical bivariate Gaussian kernel function, which transforms the point location of every volcanic event into a 2D probability density function (e.g., Connor et al., 2018). Elliptical kernels are well suited to situations where tectonism exerts a strong control on volcanism (e.g., rift settings or arcs) (e.g., Connor et al., 2018; Bertin et al., 2019). To obtain the dominant ellipse axes and orientations, the bandwidth matrices have to be calculated first (e.g., Connor et al., 2018). Amongst the many methods to estimate a bandwidth matrix, the $\hat{H}_{PLAMSE}$ selector (PI: plugin; AMSE: asymptotic mean squared error) was used in this case study due to its fast convergence to an optimal bandwidth (see Duong, 2005; Bertin et al., 2019). The influence of other bandwidth selectors on the kernel are addressed in the Discussion section.

MatHaz generates as many probability density functions as volcanic events, and groups them according to their age into different time bins. Here 36 time bins of 1 Myr each, spanning the range 36 Ma to Present, were defined in order to optimize simulation times. This required calculating the cumulative number of events per vent type for as long the vent was active with the restriction that no extra eruptions were added if the activity lasted more than 1 Myr. Thus, a rate of 44.33 (+0.78/−0.81) events per 1 Myr per polygenetic vent was obtained after extrapolating the eruption rates calculated above (see From Vents to Events). Similarly, a rate of 25.17 (−0.57/+0.56) events per 1 Myr was calculated for each polycyclic vent, and monogenetic vents were assigned a single event each. Finally, each vent was assigned to the time bin(s) representative of its period of activity (Supplementary Material S4).

For the temporal analysis it was assumed that future events will have a volume of 0.1 km³, with a mean recurrence rate of 45.77 events/10 kyr (see From Vents to Events). Spatial and temporal data were assumed to be independent (e.g., Connor et al., 2013), so multiplying them together was used to obtain the spatio-temporal matrix. The spatio-temporal probability at present was calculated based on the last 36 Myr record of volcanic activity following an age-weighting procedure using an exponentially decreasing function, with higher (lower) weights assigned to younger (older) volcanoes (for example, 10 Myr-old vents contribute ~4·10⁻⁵ whereas 20 Myr-old vents contribute ~2·10⁻⁹ to the final spatial density map; see Supplementary Material S7 for a detailed explanation on this). The probability of at least one volcanic event within a specific area (in km²) for a specific forecasting time interval (in years) was obtained by integrating the spatio-temporal matrix over a finite region. In order to work with a cumulative probability as close to 1 as possible (that is, at least one volcanic eruption is expected in the time interval of interest), a period of 10,000 years was deemed as a suitable forecasting time interval. Such a long time interval also eliminates the differences that arise when other event volumes are considered. This is further addressed in the Discussion section.

Volcanic Hazards
The most relevant volcanic hazards in the study area are: pyroclastic density currents, ballistics, lava flows (including lava domes), debris flows, and tephra fallout. The near-vent environment, with hazards associated with vent opening, short-lived phreatic explosions and toxic gas emissions, which are extremely difficult to model, are expected to be included in the ballistic ejecta zone, deemed here as an approximation of the proximal hazard zone. We did not model debris avalanches because it is unclear whether the debris avalanche deposits exposed in the study area were generated during an eruption or not (e.g., van Wyk de Vries et al., 2001; Norini et al., 2020; Rodriguez et al., 2020).

MatHaz uses an energy cone model for pyroclastic density currents (Malin and Sheridan, 1982), a two-dimensional analytical model for ballistic projectiles (Bertin, 2017), a flow inundation model for lava flows and debris flows (Connor et al., 2012), and a two-segment linear approximation for tephra fallout (Fierstein and Nathenson, 1992) (see Supplementary Material S7 for in-depth explanations about each technique). The input parameters of each model (e.g., volumes, thicknesses) were extracted from the Bertin et al. (2022) database and modeled as normal, lognormal or uniform distributions (Supplementary Figures S1–S5). The input parameters of each simulation were randomly sampled from these distributions. We ran simulations for as many different source locations as possible, which were convolved with the probabilities of a new eruption at those source...
locations within the time interval of interest (cf. Bertin et al., 2019). Results were then summed and normalized to obtain the probabilistic zoning for each volcanic hazard. We repeated this exercise ten times for each volcanic hazard to compare the outputs for the same inputs, and found that probabilities above $5 \times 10^{-5}$ can be confidently defined (Supplementary Figures S6–S9). Finally, we combined the five modeling results into a single map (cf. Javidan et al., 2021) with an even weighting for each volcanic hazard. The choice of equal weighting was made in the absence of additional evidence to define specific relative weights for the different volcanic phenomena considered in our case study, although the interested reader can assign other weights (e.g., based on expert knowledge) using the files provided in Supplementary Material S5. The limitations of the code are addressed in the Discussion section.

RESULTS

Spatial Analysis
Considering the $10^{-6}$ (events/km²) probability isocontour as a conservative boundary of volcanism at any time, most of the volcanic activity in the region is represented by a well-defined magmatic arc and small zones of back-arc activity (Figure 3A). In addition, we delineate five regions of relatively high spatial probability of future volcanic activity (above $10^{-5}$ events/km²), named here as clusters, which can be more clearly recognized if the spatial probabilities are shown following a cumulative scale (Figure 3B). These regions are:

1) Láscar cluster is N-S oriented and includes the Láscar, Colachi, Acamarachi, Chiliques, and Caicinchque stratovolcanoes, the Alítar maar, the Chascón and Áspero domes, and the Chalviri and Puntas Negras volcanic chains.
2) Socompa cluster is roughly circular and includes the Socompa stratovolcano, the Púlar-Pajonales volcanic complex, and numerous distributed lava flows and cinder cones around Socompa, such as Negros de Aras/El Negrillar and Aguas Delgadas.
3) Lazufre cluster is elliptical with a NNE-SSW axis, and includes the Lastarria, Cordón del Azufre, and Cerro Bayo stratovolcanoes.
4) Incahuasi cluster is the largest of the five and includes several stratovolcanoes and volcanic complexes that straddle the Chile-Argentina border at around ~27°S, such as Ojos del Salado, Tres Cruces, El Muerto, El Fraile, Tipas, Incahuasi, Falso Azufre, Sierra Nevada, El Cóndor, and Peinado.
5) The Antofagasta cluster is the most diffuse and the only one located in the back-arc. It includes the Cerro Blanco volcanic complex and several monogenetic mafic centres and isolated
lava flows, such as Carachipampa, Pasto Ventura, and those in the Antofagasta de la Sierra region.

Spatio-Temporal Analysis

Two spatio-temporal probability maps, generated for time intervals of 10 and 1,000 years (Figures 4A,B), show broadly similar areas of event location, with cumulative probabilities of ~5% and ~99%, respectively. The cumulative probability curve over time (Figure 5) indicates that a 50% probability of at least one 0.1 km^3 eruption occurring anywhere in the study area is estimated within ~150 years after the last 0.1 km^3-event. Cumulative probability curves constructed using other volumes are shown in Supplementary Figure S14 and are further discussed in the next section.

Probabilistic Volcanic Hazard Analysis

The probabilistic hazard map for pyroclastic density currents (Figure 6A) was obtained by using the energy cone model to forecast runout from individual volcanoes. The shape of the probability isocontours for pyroclastic density currents generally mimics the spatial distribution of volcanism due to the simple modeling approach. The lowlands within <20–30 km from the most active volcanic regions (cf. Figure 6F) show the highest probabilities for pyroclastic density currents, of which the maximum (~10^-4 pyroclastic density currents per km^2 per 10 kyr) occurs along the northern flanks of Ojos del Salado and El Muerto volcanic complexes (Incahuasi cluster). The pyroclastic density currents related to back-arc volcanic centres, such as around the Cerro Blanco volcanic complex and the Tuzgle stratovolcano, have probabilities up to 1.5 orders of magnitude lower than those estimated for the volcanic arc.

The probabilistic hazard map for ballistic projectiles (Figure 6B), which also encompasses the near-vent hazards, highlights the five volcanic clusters identified during the spatial analysis (cf. Figure 6F). In this map, probabilities rapidly decrease away from the main loci of activity, and just like the pyroclastic density currents, attain their maxima in the Incahuasi cluster. The lava flow and debris flow probabilistic hazard maps (Figures 6C,D, respectively) highlight the influence of topography. High probabilities for lava flows (~10^-4 lavas per km^2 per 10 kyr) occur within depressions or valleys, usually at <5–15 km from the summits of young stratovolcanoes. Despite a similar general appearance in the figure scale, the debris flow probabilistic hazard map (Figure 6D) highlights twice as many small ravines and catchments compared with the lava flow map. This difference is due to the lower aspect ratio (higher dispersal) of debris flows in general, as indicated by their modeling parameters. Consequently, high probabilities for debris flows (~10^-4 debris flows per km^2 per 10 kyr) extend up to 30–40 km from the major summits in the area. The highest probabilities for lava flows and debris flows (>10^-4) occur...
between the Ojos del Salado and Falso Azufre volcanic complexes, and between the El Cóndor and Peinado stratovolcanoes (Incahuasi cluster). Also highlighted by this analysis are moderate probability (~10^{-5}) zones running along the main river valleys that drain the Puna plateau towards the foreland.

Tephra fallout probabilities (normalized thickness per km^2 per 10 kyr; Figure 6E) are greatest in the prevailing downwind areas to the east of the main volcanic vents. The individual influence of each volcanic cluster (cf. Figure 6F) is well defined at probabilities >10^{-5} (cumulative probability of ~50%), but they merge at probabilities <10^{-5.5}. The extent of the 10^{-5.5} probability east of the main arc clusters are due to the influence of the back-arc Antofagasta cluster, as well the Tuzgle stratovolcano and nearby monogenetic centres. The Incahuasi cluster generates the highest probabilities for tephra fallout (~10^{-4.5}) between the El Fraile and Falso Azufre volcanoes.

**Integrated Hazard Map**

Integrating the individual volcanic hazards with an even weighting (Figure 7A) shows that most of the late Cenozoic volcanism is included within the 10^{-7} probability isocontour (defined here as the probability that an area of 1 km^2 is impacted by any volcanic hazard in a 10 kyr time frame). This region defines an overall ~700 km-long and ~300 km-wide NNE oriented ellipse. At higher probabilities, patterns are more complex. The 10^{-6} isocontour includes most Quaternary volcanic products; that of 10^{-5.5} encloses most late Pleistocene-Holocene products, while all volcanic clusters are included within the 10^{-5} isocontour. The 10^{-4.7} isocontour defines the 50% boundary, which means that if a given volcanic phenomenon occurs somewhere, there is a 50% chance it will affect the area enclosed by that isocontour. Probabilities >10^{-4} are only found in very small regions related to the Láscar, Lazufre and Incahuasi volcanic clusters. The highest probability calculated, 10^{-3.5}, occurs midway between the Ojos del Salado and Falso Azufre volcanic complexes, in the Incahuasi cluster.

Classifying probabilities as percentiles (Figure 7B; see Supplementary Figures S10–S13 for zoomed-in excerpts that highlight each volcanic cluster), three relative hazard zones are defined: high (0%–25%; >10^{-4.4}), moderate (25%–50%; 10^{-4.7}–10^{-4.4}), and low (50%–75%; 10^{-5.2}–10^{-4.7}) hazard. This means that, for instance, in the high hazard zone, an overall 25% of the volcanic hazard occurs in areas enclosed by the 10^{-4.4} probability isocontour. Whilst these bins are relatively arbitrary, our choice was guided by the desire to only show three zones (rather than many); in practice the interested reader can select other divisions by using the files provided in Supplementary Material S5.

**Coupling the Probabilistic Volcanic Hazards Assessment With Exposure Information**

The integrated volcanic hazard map shown in Figure 7A was used to calculate the relative hazard of the 692 settlements in the region (from INDEC, 2010; INE Chile, 2017) (see Table 1 for the
Most of the 10 settlements with the highest relative hazard of being impacted by a volcanic phenomenon are small villages/hamlets, administration buildings, mountain camps or mining camps. Touristic villages Antofagasta de la Sierra, El Peñón, and La Poma, rank 14th, 15th and 18th, respectively. More populous touristic towns such as San Antonio de los Cobres, in Argentina, and San Pedro de Atacama, in Chile, rank 23rd and 62nd, respectively. The three highest ranked provincial capitals are Salta (177th), San Salvador de Jujuy (219th), and San Miguel de Tucumán (353rd), in Argentina. For Chile, the three highest ranked cities are El Salvador (357th), Diego de Almagro (433rd), and Calama (642nd). In addition, the probability of a specific location being affected by any volcanic phenomenon (calculated using the integrated volcanic hazard map shown in Figure 7A) can be plotted against the spatial probability of a future volcanic event occurring directly at that location (calculated using the spatial probability map shown in Figure 3A). Figure 8 shows this exercise repeated for the 692 settlements in the region, using logarithmic scales on both axes to illustrate order-of-magnitude changes. Some settlements show low probabilities of hosting a volcanic event, but higher probabilities of being affected by any volcanic phenomenon (especially tephra fallout) (e.g., Tinogasta). Results shown in Table 1 and Figure 8 could be useful for prioritizing hazard mitigation actions, or more detailed location-specific risk analysis, such as that of Reyes-Hardy et al. (2021).
FIGURE 7 (A) Integrated quantitative volcanic hazard map, constructed by adding each probability map (Figures 6A–E), weighted evenly. Legend shows the probability of a 1 km² area to be affected by any of the five volcanic phenomena during the next 10 kyr, shown as probability isocontours on a logarithmic scale to illustrate order-of-magnitude changes. Light blue horizontal line drawn in the legend depicts the 50% boundary, which means that if a volcanic phenomenon occurs somewhere, there is a 50% probability it will affect the area enclosed by that boundary. (B) Integrated volcanic hazard map, based on the results obtained in (A), depicting three cumulative probability intervals: 0%–25%, 25%–50%, and 50%–75%, interpreted as relatively high, moderate, and low hazard regions, respectively. Zoomed-in excerpts highlighting some key regions are provided in Supplementary Figures S10–S13. The political border between Chile, Argentina and Bolivia is depicted for reference (blue line). ASTER Digital elevation model used as a background image follows a greyscale colour scheme: the whiter the pixel, the higher the elevation.

TABLE 1 The 20 settlements most likely to be affected by any volcanic phenomenon in case an eruption occurs in the study region. For the complete list of settlements analysed, the reader is referred to Supplementary Material S6.

| Locality                     | Country | Latitude (°) | Longitude (°) | Spat. Proba | Haz. Probb |
|-----------------------------|---------|--------------|---------------|-------------|------------|
| Ojos del Salado             | Chile   | −26.931      | −68.592       | 6.1·10⁻⁵    | 7.5·10⁻⁵   |
| Las Grutas                  | Argentina | −26.911      | −68.131       | 9.0·10⁻⁷    | 5.3·10⁻⁵   |
| Talabre                     | Chile   | −23.316      | −67.886       | 1.1·10⁻⁶    | 2.9·10⁻⁵   |
| Tilomonte                   | Chile   | −23.793      | −68.108       | 3.2·10⁻⁶    | 2.7·10⁻⁵   |
| Socaire (village)           | Chile   | −23.595      | −67.887       | 9.5·10⁻⁶    | 2.2·10⁻⁵   |
| Socaire (customs office)    | Chile   | −23.824      | −67.442       | 4.6·10⁻⁷    | 2.1·10⁻⁵   |
| Las Papas                   | Argentina | −26.987      | −67.782       | 5.2·10⁻⁸    | 1.8·10⁻⁵   |
| Socomba                     | Chile   | −24.451      | −68.289       | 7.5·10⁻⁸    | 1.8·10⁻⁵   |
| El Laco                     | Chile   | −23.863      | −67.491       | 6.7·10⁻⁸    | 1.7·10⁻⁵   |
| Peine                       | Chile   | −23.685      | −68.058       | 2.7·10⁻⁸    | 1.6·10⁻⁵   |
| Soncor                      | Chile   | −23.330      | −67.932       | 6.4·10⁻⁶    | 1.5·10⁻⁵   |
| Camar                       | Chile   | −23.405      | −67.958       | 6.8·10⁻⁶    | 1.2·10⁻⁵   |
| Punta del Agua              | Argentina | −27.210      | −67.731       | 3.0·10⁻⁹    | 1.2·10⁻⁵   |
| Antofagasta de la Sierra    | Argentina | −26.060      | −67.406       | 9.1·10⁻⁶    | 1.1·10⁻⁵   |
| El Perón                    | Argentina | −26.478      | −67.263       | 1.6·10⁻⁸    | 1.1·10⁻⁵   |
| Toconao                     | Chile   | −23.193      | −68.006       | 1.9·10⁻⁸    | 1.0·10⁻⁵   |
| Los Balverdi                | Argentina | −28.276      | −67.107       | 5.6·10⁻¹⁶   | 1.0·10⁻⁵   |
| La Poma                     | Argentina | −24.712      | −66.199       | 5.7·10⁻⁹    | 9.6·10⁻⁶   |
| Fiambalá                    | Argentina | −27.856      | −67.606       | 1.4·10⁻¹¹   | 9.1·10⁻⁶   |
| Medianitos                  | Argentina | −27.522      | −67.580       | 1.8·10⁻¹⁰   | 6.9·10⁻⁸   |

aProbability of a volcanic event taking place in the settlement over the next 10 kyr.
bProbability that the settlement is affected by any volcanic phenomenon over the next 10 kyr.
examining settlements in relation to Guallatiri volcano (18.5°S/69°W).

**DISCUSSION**

**Volcanic Clusters, Magma Bodies and Thermal Anomalies**

The volcanic clusters identified in this work can be indicative of subvolcanic processes that could lead to large-volume volcanism in the future (cf. George et al., 2016). In our study region, geophysical studies have identified large-scale crustal anomalies related to shallow (~2–10 km) partial melt bodies (Ward et al., 2013, Ward et al., 2017; Pritchard et al., 2018) (Figure 9). These include the Altiplano-Puna Magma Body (~21–24°S; Chmielowski et al., 1999; Ward et al., 2014; Pritchard et al., 2018), which is a ~500,000 km$^3$ mid-crustal body with an average 15%–22% partial melt (Schilling et al., 2006; Comeau et al., 2016; Pritchard et al., 2018; Spang et al., 2021). The Láscar volcanic cluster is located in its southern portion, also recognized by Díaz et al. (2012) using magnetotelluric data. The Láscar cluster may therefore represent a subdomain of the broader magma body with higher melt fractions. The Socomba cluster identified here lies outside the known boundaries of the Altiplano-Puna Magma Body to the south and may either represent a previously undocumented southern extension of it, or some other source anomaly such as a small zone of recent lithospheric delamination (Beck et al., 2015; DeCelles et al., 2015; Schoenbohm and Carrapa, 2015).

Further south, the Southern Puna Magma Body (~25–27.5°S; Bianchi et al., 2013) comprises at least four separate mid-to-upper crustal magma bodies: Lazufre, Incahuasi, Cerro Galán, and Incapillo-Bonete (Figure 9), with a cumulative volume of ~90,000 km$^3$ (Delph et al., 2017, Delph et al., 2021; Ward et al., 2017). The Lazufre and Incahuasi volcanic clusters overlap with the eponymous magma bodies, although the latter volcanic cluster is much larger than the magma body identified. The Incapillo-Bonete and Cerro Galán magma bodies do not appear to be linked to volcanic clusters. The Cerro Galán magma body is related to the eponymous caldera, which records at least ~4 Myr of ignimbrite-forming eruptions (Folkes et al., 2011; Kay et al., 2011), yet there has been no late Pleistocene-Holocene volcanic activity at this centre. The Incapillo-Bonete magma body is deeper than the others (~30–50 km; Goss et al., 2011; Ward et al., 2017) and may need to grow or rise before large-volume volcanism starts. The Antofagasta volcanic cluster identified here is, apparently, not associated with a previously defined magma body nor mid-to-upper crustal region of partial melt (Ward et al., 2017), although it is an actively deforming region, hosting the only known active caldera in the CVZ (Cerro Blanco) (Pritchard and Simons, 2002; Henderson and Pritchard, 2013; Vélez et al., 2021; de Silva et al., 2022).
Surface thermal anomalies have been detected through infrared satellite imagery at nine volcanoes: Licancabur, Alítar, Lásca, Chiliques, Púlar-Pajonales, Lastarria, Sierra Nevada, Falso Azufre, and Ojos del Salado (Jay et al., 2013). Of these, all but Licancabur are either on or near the edges of the volcanic clusters identified here, and four of them (Alítar, Lásca, Lastarria and Ojos del Salado) host permanent fumarolic activity. Volcanoes analysed by Jay et al. (2013), however, were only those selected from the Siebert and Simkin (2002-2010) catalogue, thus some volcanoes with the potential for thermal anomalies (i.e., with crater lakes and/or sulphur deposits) were not analysed (e.g., Chalviri, Tres Cruces, Cerro Blanco, El Fraile, El Muerto). Some of those volcanoes should be the focus of future field studies as they lie within the clusters identified here.

**Arc vs. Back-Arc Distinction**

In the Puna plateau, there was a kinematic shift from compression to strike-slip and minor extension at 12–10 Ma (Marrett et al., 1994; Schoenbohm and Strecker, 2009; Daxberger and Riller, 2015). The oldest back-arc mafic lavas in this region to have erupted since this shift have been dated at ~7.3 Ma (Risse et al., 2008) (although other undated lavas might be older), whereas the youngest ones are some extremely well preserved lava flows and cinder cones in the Antofagasta de la Sierra region (Risse et al., 2008; Fernández-Turiel et al., 2021). Based on the spatial probability map of volcanic activity (Figure 3A), the current volcanic back-arc can be delineated within the $10^{-54}$ probability isocontour (Figure 10). Matching the western boundary of the back-arc with the eastern boundary...
of the volcanic arc, these two regions can be compared to analyse their respective contributions to the volcano by integrating the probabilities enclosed by each region. Thus, in terms of the probable locations of future volcanic activity, the arc contributes 88% and the back-arc 12%, whereas in terms of areas likely to be affected by future volcanic activity, the back-arc contributes 81% and the arc 19%. We therefore conclude that the arc is seven times more likely to host volcanic activity than the back-arc, and four times more likely to be affected by volcanic phenomena than the back-arc.

Sensitivity Analysis and Limitations of Our Study

The hazard assessment conducted in this study follows a sequential methodology that relies upon some assumptions. For instance, the estimation of volcanic events per vent type depended on the assumption of a fixed volume of 0.1 km$^3$ per event. Even though this volume can be deemed reasonable for the purposes of our application, considering that the large majority (>90%) of explosive volcanic eruptions in the world are of small-to-moderate size (VEIs of 0–3), and that ~20% of Holocene eruptions have VEIs of 3–4 (e.g., Siebert et al., 2010; Papale, 2018), it is necessary to test the influence of other volumes on the estimated eruption rates. We therefore tested order-of-magnitude differences in volume: 0.01 and 1 km$^3$. A volume of 0.01 km$^3$ led to a polygenetic eruption rate of 4.5855 (~0.0801/0.0802) events per 10 kyr and an estimated 183.23 (~10.21/10.90) events for all post-10 ka active vents. For a volume of 1 km$^3$, the polygenetic eruption rate was 0.0315 (~0.0008/0.0008) events per 10 kyr and the eruption rate for all post-10 ka active vents was 32.06 (~0.80/0.84) events. When the cumulative probabilities calculated from these eruption rates are projected into the future and plotted alongside the cumulative probabilities used throughout our case study (Supplementary Figure S14), we see that at large (>2,000 years) forecasting time intervals, all curves have attained cumulative probabilities very close to 1. Thereby, the forecasting interval of 10 kyr used during our hazard assessment can be considered as a reasonable and conservative value because the cumulative probabilities sum to near unity irrespective of the event recurrence rate.

Two main sources of uncertainty can affect the spatial probabilistic analysis (Bertin et al., 2019). The first refers to how bandwidth selectors can change the shape of the probability isocontours in the spatial probability map. Here, besides the $\hat{H}_{PLAMSE}$ selector used in our case study, we also tested the influence of two other commonly used bandwidth selectors, $\hat{H}_{PLSAMSE}$ (PI: plug-in; SAMSE: sum of asymptotic mean squared errors) and $\hat{H}_{SCV}$ (SVC: smoothed cross validation) (see Duong, 2005; Bertin et al., 2019), on the spatial probability map shown in Figure 3A. Results (Supplementary Figure S15) show that the largest discrepancies are found towards the lowest probability isocontours (<$10^{-10}$ in our case), so the location and extent of the spatial clusters, which were defined at the highest probabilities, do not change when other bandwidth selectors are used. The second source of uncertainty is related to the distribution of vents into different time bins according to their modeled ages (ages sampled from normal distributions). Here we compared the distribution of vents used during our analysis to the range of distributions obtained after repeating the vent-grouping exercise $10^3$ times. If we plot the minimum and maximum number of vents grouped within each time bin (Supplementary Figure S16), we see differences no higher than ~3% compared to the number of vents per time bin used during our study. Further, if we extract the minimum and maximum number of vents for the most recent time bin (<1 Ma) and model the ellipses of their respective bandwidths (calculated using the $\hat{H}_{PLAMSE}$ selector), it is found that the orientation (i.e., azimuth) of the ellipses differ by less than 0.3% compared to the orientation of the ellipse used during the spatial analysis. Similarly, a minor difference (<0.4%) is obtained when the areas of these ellipses are compared to each other. In conclusion, other bandwidth selectors and different vent-grouping age distributions do not affect the spatial probability map to any significant extent.

MatHaz has been developed as a tool to help deal with the real and practical complications of assessing hazard for large volcanic areas (Bertin et al., 2019). In order to achieve this task within reasonable simulation times, and unlike other hazard simulation
tools, MatHaz has not been developed to model the physics of volcanic eruptions and their products, or to model the details of specific eruptive scenarios (Bertin et al., 2019). This strategy focuses on simplified models that are easy to manage and replicate and quick to execute, acknowledging that by meeting these conditions model resolution is lost. Nevertheless, even the simplest models considered in MatHaz have their own assumptions. For example, tephra fallout modeling was performed by drawing elliptical isopachs assuming a deposit that thins exponentially away from the source, an approach that is not necessarily true based on well-documented eruptions (e.g., Bonadonna and Costa, 2012; Kawabata et al., 2013). Debris flows and lava flows are modelled based on a simple ‘flow-routing’ code, so an important limitation is that the hazard areas will always follow the thalweg of the valley and will never overbank.

The simulation times of each model tested are also strongly influenced by the pixel size, which determines the level of spatial detail of the assessment. In our case study application, a pixel size of 1,000 m was used, which is ~0.1% of the longest side of the study area. This proportion was deemed adequate by Bertin et al. (2019) after testing the pixel size/computational efficiency trade-off for different pixel sizes, as any improvement on the spatial resolution of the assessment was far more computationally intensive but with little reward, in that the shapes of the main probability isocontours did not differ significantly for pixel size/study area ratios lower than 0.1%.

For the modeling methodology, it was assumed that every volcanic phenomenon had its source in a single pixel of fixed size. This is an oversimplification, especially for lava flows, since these can also start erupting from a fissure, which may or may not evolve into a single vent (e.g., Valentine and Gregg, 2008). In MatHaz, simulation of fissure eruptions adds a major complexity to the model due to the many assumptions about which additional source vents will have to be considered. In spite of these limitations, for a future application of MatHaz a potential user can edit the code to include other modeling algorithms and simulation tools. Another possibility might be running more sophisticated numerical models for those regions where the limitations, for a future application of MatHaz a potential additional source vents will have to be considered. In spite of these limitations, for a future application of MatHaz a potential user can edit the code to include other modeling algorithms and simulation tools. Another possibility might be running more sophisticated numerical models for those regions where the limitations, for a future application of MatHaz a potential user can edit the code to include other modeling algorithms and simulation tools.

Summarizing, our approach is, at least, effective in providing a first-order assessment of volcanic hazards for large and complex volcanic regions. In addition, even though we acknowledge that our results are mostly valid at a large (regional) scale, they can be used to guide targeted hazard assessments at more detailed scales.

**Improving Future Hazard Assessments With Structural, Geophysical, and Geochronological Datasets**

Probabilistic volcanic hazard assessments can be improved by including robust volcano-structural, seismic and thermal datasets at more detailed scales (e.g., Bartolini et al., 2014; Bertin et al., 2019). This information is however still patchy and incomplete for the area considered in the present study. For instance, apart from two volcanoes (Láscar and Lastarria), there is no permanent seismic network in the region (Amigo, 2021; Aguilera et al., 2022). Other isolated seismicity has been considered tectonic, but could have magmatic influences; for instance, Mulcahy et al. (2014) identified two brief, shallow (<5 km) swarms east of Falso Azufre and at Cerro Galán.

There is a large catalogue of faults, fractures, fissures and lineaments for this area (5,887 structures; extracted from the Bertin et al., 2022 database) (Supplementary Material S4), although these are mainly mapped in basement units, with structures unrecognised beneath most of the volcanic centres. To gain insight into how the structures can influence a future hazard analysis, we tested the incorporation of this patchy structural data into the probabilistic spatial assessment following the methodology described in Bertin et al. (2019). Results (Figure 11) show that even conservative weightings on structures (~20%) distort the probability isocontours towards regions far away from known volcanism. Hence, it appears that homogenous and high-resolution structural information (e.g., Becerril et al., 2013; Bartolini et al., 2014; Bertin et al., 2019; Bevilaqua et al., 2021) is required before it becomes useful, especially in order to discriminate between those structures that have interacted with ascending magmas in the past, and those that have not. In other words, regional-scale structural data do not seem to improve the assessment.

Data indicative of persistent thermal anomalies and/or zones of active surface deformation could also be used to improve future hazard assessments, although homogeneity of data coverage must be considered. Low enthalpy geothermal springs have been mapped in Chile (Hauser, 1997) and Argentina (Pesce and Miranda, 2003), although these are mostly in places accessible by road. Fumarolic activity at some volcanoes (e.g., Tassi et al., 2009, Tassi et al., 2011; Aguilera et al., 2016; Chiodi et al., 2019), together with geological evidence of young (i.e., Quaternary) phreatomagmatic activity at others (e.g., van Alderwerelt, 2017; Filipovich et al., 2019, Filipovich et al., 2020; Ureta et al., 2020, Ureta et al., 2021), most of which fall within the volcanic clusters identified here, should be coupled with high-resolution thermal surveys to improve the spatial coverage of potential current magmatic anomalies. On the other hand, there are long-term ground deformation anomalies identified in InSAR time series, tilt meter data, GPS/GNSS permanent stations and ground surveying at some active volcanoes (Pritchard and Simons, 2002; Henderson and Pritchard, 2013; Amigo, 2021; Vélez et al., 2021; Aguilera et al., 2022). This information could be used to guide high-resolution geodetic studies, following on from some targeted surveys that have been conducted at Cerro Blanco (Di Filippo et al., 2008; Brunori et al., 2013; Chiodi et al., 2019; Lamberti et al., 2020; Vélez et al., 2021).

Improvement of the geochronological data will definitely help constrain the periods of activity of individual volcanic vents. Summing all available data into a geochronological database (Bertin et al., 2022) is robust over long time periods (10s of Myr), or better in some areas, but there is a paucity of data to effectively characterise the late Pleistocene-Holocene volcanic record, especially when many morphologically young volcanic deposits remain undated. A desirable future goal would be to gain enough new accurate and high-precision age determinations to improve the temporal resolution of this modeling approach ten-
fold (i.e., from 1 Myr to 100 kyr analysis windows). This will require targeting well-preserved recent volcanic deposits with state-of-the-art dating techniques to have the best chance of defining robust probability density functions for shorter time intervals, especially because the youngest time window is the one that contributes the most to the spatial probability map.

**Recommendations for Future Work**

Our study area is too large to expect a uniform high-resolution assessment of the volcanic eruption record. However, targeted analyses on particularly active parts of the region could lead to a much improved hazard assessment. We suggest high-resolution studies on the Puntas Negras and Chalviri volcanic chains (Láscar cluster), the Socompa and Púlar-Pajonales volcanoes (Socompa cluster), as well as the many volcanoes within the Incahuasi cluster (e.g., Ojos del Salado, El Fraile, Tipas, Peinado). The Incahuasi region shows the highest event probabilities (Figures 3, 4), and has been addressed in only a few studies (González-Ferrán et al., 1985; Baker et al., 1987; Mpodozis et al., 1996; Gardeweg et al., 2000; Grosse et al., 2018, Grosse et al., 2020, Grosse et al., 2022; Naranjo et al., 2019). In addition, at least two <1 ka BP tephras mapped in Argentina (Fauqué and Tchilinguirian, 2002; Hermans and Schellenberger, 2008), were likely erupted from the Incahuasi cluster (Sampietro-Vattuone et al., 2018, Sampietro-Vattuone et al., 2020; Fernández-Turiel et al., 2019). Additional permanent seismic monitoring networks should also be considered at Socompa, Tres Cruces, Ojos del Salado, and Cerro Blanco.

We recommend conducting a comprehensive evaluation of the erupted volume and volcanic eruption rates for this arc segment, as it could allow for an in-depth reconstruction of the spatio-volumetric-temporal evolution of volcanism in the region (e.g., how volcanism in terms of intensity and magnitude has varied over time). To achieve this goal, however, will require the development of a model that explicitly integrates the spatial, temporal, and volumetric components of hazard. Sophisticated spatio-temporal (Bebbington, 2013) and spatio-volumetric (Bebbington, 2015) hazard models have been developed for distributed volcanic fields, but their applicability to arc segments is yet to be tested. The inclusion of a volumetric component into a well-tested spatio-temporal model may shed further light on how much each volcanic cluster identified here contributes in terms of erupted volume. This will, in our opinion, highlight the Antofagasta cluster, particularly due to the presence of the Cerro Blanco volcanic complex, which has long been recognized as the source of at least two eruptions on the scale of the largest eruptions to have occurred on Earth since the late Pleistocene (Báez et al., 2015; Báez et al., 2020a; Báez et al., 2020b; Fernández-Turiel et al., 2019; de Silva et al., 2022).

**CONCLUSION**

We developed a long-term probabilistic volcanic hazard assessment of the Chilean-Argentinian segment of the Central Volcanic Zone of the Andes (~22.5–28°S). Through this we
provide a first-order framework on which to build a detailed understanding of arc-scale/regional volcanic hazard. We recognize five regions of high spatial probability of future volcanic activity: Láscar, Socompa, Lazufre, Incahuasi, and Antofagasta. The largest clusters (Láscar, Lazufre and Incahuasi) correlate well with geophysical evidence of mid-crustal partial melt bodies, but the smallest two (Socompa and Antofagasta) do not. The Socompa and Antofagasta clusters should therefore be considered for further targeted thermal and other geophysical studies. We estimate a probability of 50% for an eruption of the scale of the Láscar 1993 event (0.1 km$^3$) by the year ~2150. Our results suggest an ~80% probability that the next 0.1 km$^3$ eruption will occur from within one of the five volcanic clusters. The lack of a robust late Pleistocene-Holocene eruption record in this area is still a major hindrance for accurate short-term hazard assessments.

Five volcanic phenomena (pyroclastic density currents, ballistic projectiles, lava flows, debris flows and tephra fallout) were modeled for the whole region. Considering these collectively in an integrated volcanic hazards map, a relative hazard exposure was established for population centres in the region. This showed that the settlements with the highest relative likelihood of being affected by any of the volcanic phenomena mentioned above are small towns or similar, some of which are popular tourist destinations. Our results suggest that any major efforts towards improving volcanic hazard knowledge for this region should concentrate on improving volcanic event records within each of the five volcanic clusters identified here, especially the Incahuasi and Antofagasta clusters. In addition, examining the volcanic risk associated with the largest population centres that were defined in our exposure analysis should be pursued, especially San Antonio de los Cobres in Argentina and San Pedro de Atacama in Chile.

We suggest that the sequential methodology developed here could be tested in other volcanically active regions, as the results of such regional analysis can be used to identify areas to be prioritised for future volcanological research. Future efforts should also seek to explicitly add the erupted volume to the spatio-temporal framework in order to better understand other arc-scale processes that can impinge upon volcanic hazards.

**AUTHOR CONTRIBUTIONS**

DB conceived the study, developed methods, wrote and run all code, wrote the manuscript and prepared all relevant figures and complements. This work was supported and supervised by JL, SC, SdS, CC, and PC, who also contributed to the interpretation and edited the manuscript with several inputs from PG, WB, EB, and RC. All authors read and approved the final manuscript.

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**DATA AVAILABILITY STATEMENT**

The original contributions presented in the study are included in the article/Supplementary Material, further inquiries can be directed to the corresponding author.

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**SUPPLEMENTARY MATERIAL**

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/feart.2022.875439/full#supplementary-material

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