

Investigating the Crystal Cargo of Fiamme from the Ora Ignimbrite (Permian, Italy): Elucidating the Spatio-Temporal Evolution of a Crystal-Rich, Supereruption-Sized Magma System

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Authors

- **Genna Chiaro**

 [0000-0002-8981-7679](#) ·  [gennachiaro](#) ·  [gennachiaro](#)

Department of Earth and Environmental Sciences, Vanderbilt University

Motivation

Supereruptions, events that expel >450 km³ of magma from the subsurface, are the most extreme expressions of eruptive volcanism on Earth (Sparks et al., 2005). These catastrophic events tend to occur every ~100,000 years (Mason et al., 2004). Due to their scarcity, we are limited to studying the resulting deposits of these colossal eruptions. By investigating the evolutionary history preserved in mineral rims, unaltered volcanic glass compositions, spatial distribution of the deposits, and timescales of magmatic processes, we can inform our current hypotheses regarding the formation of large silicic magma systems that are capable of supereruption.

Our inability to obtain high-resolution subsurface images has spurred innovative methods for resolving the organization of magma in the crust. If we know the major element concentrations of the original melt and the minerals present, we can obtain temperature and pressure estimates using the thermodynamics software, Rhyolite-MELTS (Gualda and Ghiorso, 2014). Importantly, this can help determine where the magma bodies were situated in the crust prior to eruption. Major and trace element compositions of volcanic glasses can provide information on the number of magma bodies that contributed to the supereruption (Begué et al., 2014). Furthermore, both the lateral distribution of volcanic glasses and ⁴⁰Ar/³⁹Ar sanidine geochronology can provide insight on the evolution of the eruption and whether the deposit was emplaced instantaneously or had distinct volcanic pulses (Begué et al., 2014; Kay et al., 2011).

I plan to investigate the pre-eruptive conditions of a crystal-rich supereruption, the Ora Ignimbrite, in order to fill current gaps in knowledge regarding the structure and evolution of highly-evolved, crystal-rich, trans-crustal magma systems.

The following research questions are the main focus of my dissertation:

- 1. Can we use mineral major and trace element compositions to identify fiamme type and population?**
- 2. Do we observe similar histories recorded in the mineral populations? What can this tell us about magmatic processes during the assembly and eruption of large silicic systems?**
- 3. Can we observe clustering of fiamme populations in space and time? What does this suggest about the subsurface architecture of the Ora magmatic system? Are we able to identify different phases of the eruption and the tapping of separate vents?**

4. Can high-precision $^{40}\text{Ar}/^{39}\text{Ar}$ sanidine ages subdivide the Ora eruption into distinguishable volcanic pulses, thereby illuminating the episodic tempo and dynamic behavior of this very large magmatic system?

The Permian Ora Ignimbrites ($>1,200 \text{ km}^3$) are advantageous deposits for this study because glacial incision has revealed exceptional intracaldera exposures with $>1,000 \text{ m}$ of vertical relief. I plan to expand upon my M.S. work on the Ora vitrophyre horizons and develop major and trace element-based hierarchical clustering algorithms in python to identify minerals that correspond to the pre-defined fiamme types and populations identified in Chiaro (2019). A hot-spot clustering analysis (Ripley's k-function) in ArcGIS will be used to determine the distribution of fiamme types and populations throughout the entire deposit and high-precision $^{40}\text{Ar}/^{39}\text{Ar}$ sanidine ages will provide eruption ages. By integrating this project with the Rhyolite-MELTS pressures calculated in Chiaro et al., (in prep), we hope to elucidate the spatio-temporal evolution of a supereruption-sized, crystal-rich magma system.

Background

The supereruption-scale Ora Ignimbrites are products of an ignimbrite flareup resulting from subduction of the Paleotethys Ocean and the consequent formation of Pangaea (Cassinis et al., 2012). Glacial incision of the Permian Ora Ignimbrites ($>1,200 \text{ km}^3$) reveals heterogeneous intracaldera deposits with $>1,000 \text{ m}$ of relief. Less voluminous outflow was emplaced to the north and south of the caldera and is stratigraphically similar to late-erupted intracaldera units (Marocchi et al., 2008). The Ora deposits are interpreted to be low-intensity eruptions that tapped multiple magma bodies and were sourced from discrete vents (Willcock et al., 2015). Bulk rock chemistry and mineralogy suggest that caldera collapse and eruption initiated in the south and migrated northward through time (Willcock et al., 2013).



Figure 1: A map of the field location. The four fiamme types identified in Chiaro et al. (in prep) are listed with their corresponding vitrophyre locations.

Table 1: Samples collected in Fall 2017 and Spring 2019. Sample locations are shown in Fig. 1. The number of bulk tuff samples and fiamme are listed.

Samples:	Location:	Type:	Bulk Tuff:	Fiamme:
ORA 2	Outflow Vitrophyre	Vitrophyre	9	31
ORA 3	Northern Intracaldera	Bulk Tuff	1	1
ORA 4	Odorizzi Quarry	Bulk Tuff	2	50
ORA 5	Intracaldera Vitrophyre	Vitrophyre	1	30
ORA 6	Northern Intracaldera	Rhyolite Dike	4	

Samples:	Location:	Type:	Bulk Tuff:	Fiamme:
ORA 7	Base of Northern Intracaldera	Bulk Tuff	1	
ORA 8	Northern Intracaldera	Bulk Tuff	1	
ORA 9	NW Outflow 1	Biotite-Rich Fiamme		1
ORA 10	NW Outflow 2	Medium-Grained Fiamme		1
ORA 11	Castel Firmiano	Ash/Ignimbrite Contact	3	
ORA 12	Rhyolite Dike	Granite Dike	4	
ORA 13	Rhyolite Dike	Potential Fiamma		1
ORA 14	Caldera Rim Cutting Dike	Porphyritic Coarse-Grained Dike	1	
ORA 15	Caldera Rim Cutting Dike	Medium-Grained Dike	1	
ORA 16	Cava Flor Quarry	Fine and Coarse-Grained Fiamme		4
ORA 17	NW Outflow 3	Fine to Medium-Grained Fiamme		3
ORA 18	Magdalena Hike	Pseudotachylite	1	
ORA 19	Magdalena Hike @ Sculpture	Fine and Coarse-Grained Dikes	4	

Recent work investigated the pre-eruptive architecture of the Ora system with textural, mineralogical, and geochemical features of fiamme (Chiaro, 2019). Fiamme were collected from two vitrophyre horizons, an early-erupted intracaldera deposit and a late-erupted outflow deposit, in order to find well-preserved glass. These fiamme were initially grouped into four types based on crystal content and biotite content (Table 2). Glass major elements further verified the categorization scheme (Fig. 2). Tight clustering of MG and VCCR glass major element compositions suggests that the glasses are unaltered and thus represent pristine melt compositions (Fig. 2).

Table 2: The four fiamme types and their location, crystal content, maximum phenocryst size, and mafic content (Chiaro et al., in prep).

FIAMMA TYPES:	DESCRIPTION:	LOCATION	CRYSTAL CONTENT:	MAX PHENOCRYST SIZE:	MAFICS:
TYPE 1:	Very Coarse-Grained Crystal-Rich (VCCR)	Intracaldera	~40-50%	>5 mm	~2-3%
TYPE 2:	Medium-Grained (MG)	Outflow	~20-40%	1-3 mm	<0.5%
TYPE 3:	Fine-Grained (FG)	Intracaldera	~20%	≤1 mm	<0.1%
TYPE 4:	Fine-Grained Crystal-Poor (FGCP)	Outflow	~10%	≤1 mm	≤0.15%

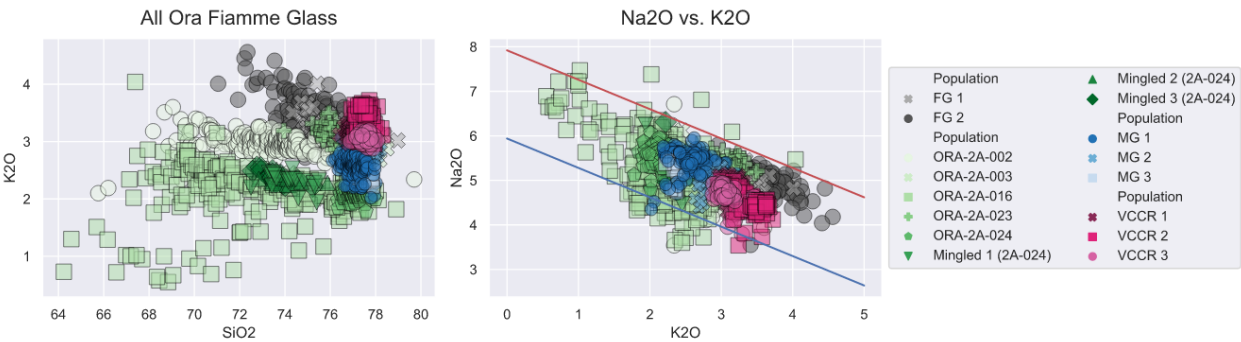


Figure 2: All fiamme major element glass compositions. (a) The tight clustering of VCCR and MG fiamme suggest that these glasses did not undergo alteration. (b) Clustering in the Na₂O vs. K₂O plot provides further evidence that the VCCR and MG glasses represent pristine melt compositions.

Glass trace elements illuminate at least three distinct populations for both the VCCR and MG fiamme, suggesting the presence of multiple magma batches within the Ora system (Fig. 3). Pre-eruptive storage pressures from the Rhyolite-MELTS (Q2F) geobarometer suggest that there were two potential scenarios for the subsurface organization of magma bodies in the Ora system (Fig. 4). Either there were two distinct crystal mush zones located at slightly different depths within the crust, or there was a continuous crystal mush present with multiple heterogeneous zones located throughout (Chiaro, 2019).

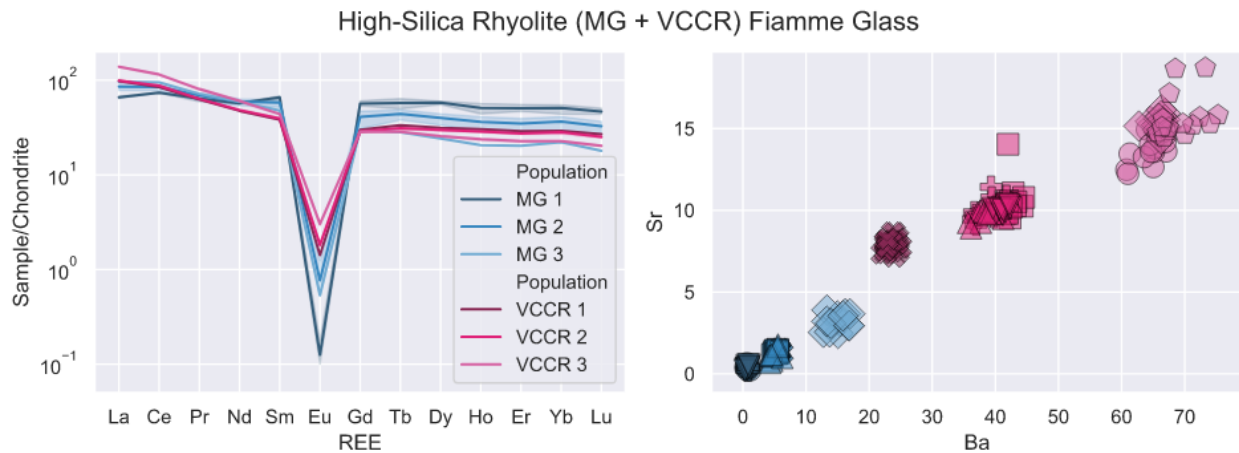


Figure 3: Rare earth element diagrams and a plot of strontium vs. barium illustrate the six discrete glass populations in the MG and VCCR fiamme.

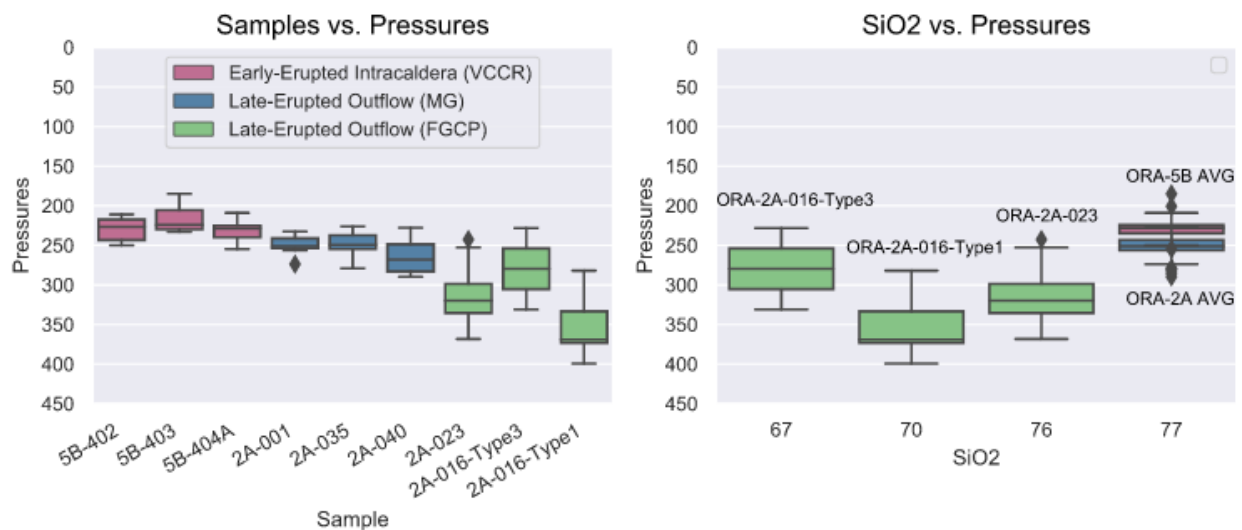


Figure 4: Different fiamme types were stored at different pressures in the crust, with a progression of erupting shallower to deeper magmas through time. Silica content vs. pressure shows variable storage depths for the FGCP melt, suggesting that there were multiple, small melt bodies located at different depths throughout the magma chamber (from Chiaro et al., in prep).

Approach



Building on the work from Chiaro et al. (in prep), we propose to use mineral compositions in order to apply this characterization scheme to the entire Ora deposit. Since unaltered glass is only present in the vitrophyre horizons, we plan to use mineral compositions to classify each fiamma. We will collect trace and major elements of plagioclase, alkali feldspar and biotite in fiamme. Preliminary work shows that biotite and plagioclase major elements can be used to differentiate intracaldera vs. outflow fiamme. Furthermore, biotite major element compositions can be used to infer both fiamme type and population (Fig. 4).

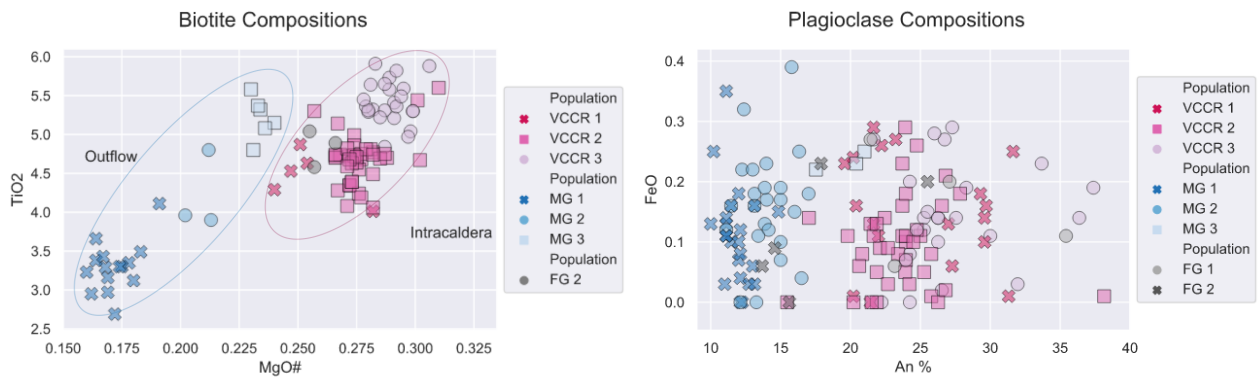


Figure 4: Biotite and plagioclase major element compositions from fiamme. Fiamme are divided into population based on hue. Both plagioclase and biotite can be used to determine fiamme type and biotite can also determine fiamme population (from Chiaro et al., in prep).

Hierarchical Clustering and Machine Learning in Python:

The foundation for this project involves obtaining mineral compositions for the entire crystal. After creating biotite and feldspar EDS maps and collecting trace element profiles, I plan to utilize a hierarchical agglomerative clustering algorithm for major elements in order to statistically determine how many clusters of minerals are present in the fiamme. Another clustering algorithm will be run using trace elements and we will compare the results. Once the clusters are identified, we can observe whether they correlate with fiamme type and population (Chiaro et al., in prep). If we see a correlation, we can confidently use mineral populations to classify fiamme throughout the Ora Ignimbrite. We can also use the mineral maps to visualize core to rim cluster evolution (Fig. 5). By visualizing where the clusters occur in each mineral and by looking at a large enough population of core to rim clusters, we can infer whether crystals experienced similar histories. This approach may reveal information regarding the magmatic processes that occurred prior to the Ora eruption (injection, decompression-driven growth, etc).

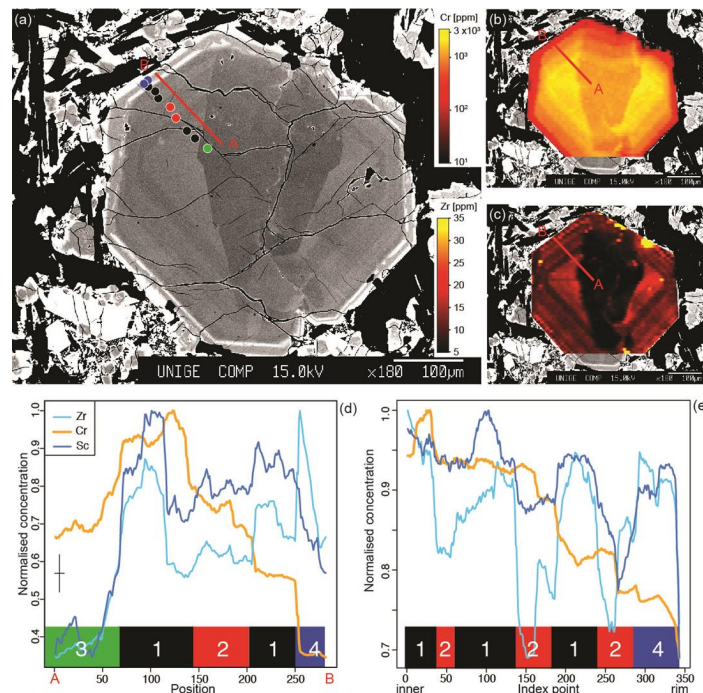


Figure 5: (A-C) BSE and EDS scans of a clinopyroxene phenocryst from the Holuhraun-Bardarbunga eruption in 2014-2015. The circles show the location of the electron microprobe analysis and the color indicates the corresponding cluster from major element analysis. (D) Trace element profiles from A-B.

The colors and numbers at the bottom indicate the corresponding cluster identified with major elements. (E) The same as D, but for another crystal that is not shown (from Luca Caricchi).

We are also interested in identifying the mineral populations that are present in the ignimbrite matrix and seeing whether this varies throughout the deposit. Machine learning algorithms are becoming widely incorporated in scientific studies, and they provide an opportunity to make tedious point-counting obsolete. By utilizing previously determined mineral clusters (major element-based clustering), we can train a machine to classify each mineral in a thin section EDS scan into its associated population. Furthermore, this automatic clustering can be applied to multiple large thin sections, allowing for an estimation of bulk eruptive material. Initial formulation will involve training a supervised machine learning algorithm to identify feldspars. If successful, we will work to develop an unsupervised neural network algorithm. The unsupervised neural network algorithm will effectively create clusters automatically via self-organizing maps and then it will tag each mineral with its identified population.

Spatial Statistics in ArcGIS:

The spatial distribution of fiamme types and populations has the potential to reveal the locations of eruptive vents that were destroyed during the eruption. Multi-distance spatial cluster analysis (Ripley's k-function) in Arc-GIS investigates the clustering of features over a defined study area. Similarly, this may provide insight on the pre-eruptive magmatic architecture. For example, if the Ora system had two laterally juxtaposed magma chambers that were tapped by different eruptive conduits, we would expect to find two separate clusters with discrete fiamme populations. Integrating spatial statistics and petrology can help us determine how magma was stored in the crust prior to eruption (single chamber, multiple chambers?) and the approximate locations of storage in the x-y plane. If we combine our previously obtained Rhyolite-MELTS data to this model, we can elucidate pre-eruptive storage conditions in three-dimensions.

Sequence Stratigraphy and Ar-Ar Age Dating:

We propose to investigate whether fiamme type and population change as a supereruption progresses. Our approach will take advantage of the previous work done by Willcock et al. (2013) in which they correlated stratigraphic logs throughout the deposit with observed changes in volcanic lithofacies for 13 intracaldera and 7 outflow stratigraphic sections. More detailed efforts to ascertain the temporal changes of the Ora eruption involved plotting the bulk rock compositions, biotite crystal content, and total crystal content as they varied stratigraphically (Willcock et al., 2015) (Fig. 6). We have identified potential horizons of interest that may signify a change in eruptive material (Fig. 6). By sampling the fiamme types present at these locations, we hope to discover whether there are gradual or instantaneous changes in the fiamme types and populations. Similar to the methodology employed in sedimentary strata with fossil assemblages, we may be able to observe fiamme "extinction horizons" representing the depletion of a magma chamber. We will also utilize this workflow on samples of the ignimbrite matrix with a machine learning algorithm to determine whether we see a change in mineral populations with time. These methods will allow us to extract relative timescales for magmatic events taking place during the Ora supereruption.

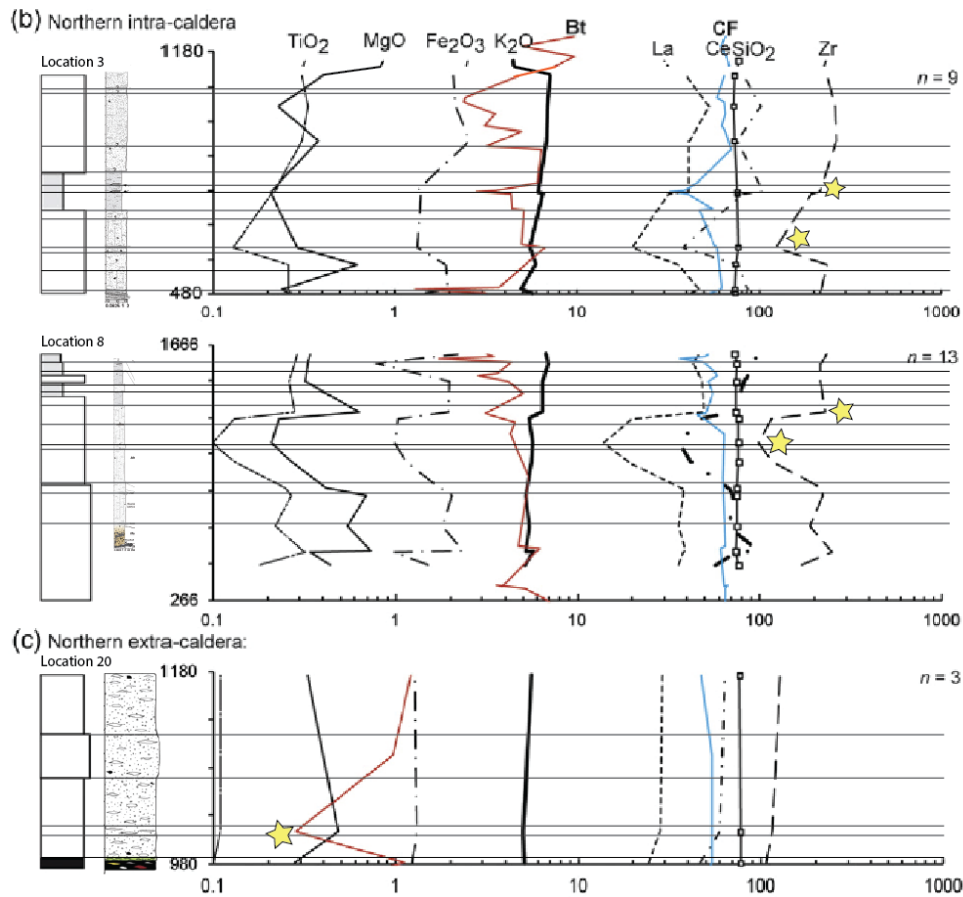


Figure 6: Bulk rock compositions, biotite crystal content (Bt), and total crystal content (CF) variation through time. The yellow stars indicate horizons of interest that could show potential changes in eruptive material. Locations 3 and 8 are intracaldera deposits and correlate with Ora 3 and Ora 8. We have yet to sample from location 20, which is just south of the Ora 2 vitrophyre unit. Figure is modified from Willcock et al., 2015.

There has not been much work done regarding age dating for the Ora Ignimbrite. Marocchi et al. (2008) obtained single zircon ages for Ora Ignimbrite of 277 ± 2 Ma for the intracaldera fill and 274.1 ± 1.6 Ma for the outflow. These suggest a time break in the eruptive record. Willcock et al. (2015) suggests that Ora experienced a two-stage eruptive history: the southern caldera collapsed first and the northern caldera followed. However, there is no evidence for paleosols or reworked contacts, suggesting that the eruptions were closely spaced in time. The zircon ages are in contrast with the inferred eruption timescale based on stratigraphic evidence. In order to determine whether the Ora eruption lasted for millions of years or was constrained to a shorter time period, we plan to obtain Ar-Ar sanidine ages. Argon is trapped upon eruption so these will provide eruptive ages. We will first determine eruption ages for the intracaldera vitrophyre and the outflow vitrophyre to see if we can resolve discrete ages. If so, Ar-Ar sanidine ages provide a methodology to “deconstruct” a supereruption into distinct volcanic pulses (Kay, 2011). Combined with our work on fiamme type distribution for a stratigraphic column, we may be able to resolve maximum timescales of magma chamber depletion for the Ora system.

Methods

Sample Collection:

The samples collected during the 2017 and 2019 field seasons are shown in figure 1 and their descriptions are listed in table 1. Currently, we have a disproportionate amount of samples from the northern intracaldera. In order to reduce the sampling bias for our spatial statistics, we will obtain

more samples from the southern intracaldera and southern outflow. Clustering analyses in ArcGIS requires at least 30 features to be reliable. In order to sample a significant number of fiamme, we will make fiamme mounts to maximize sampling potential and analysis speed. The ignimbrite matrix samples will be made into 75 x 50 mm billets.

Analytical Work:

We plan to obtain EDS scans of individual mineral grains and corresponding BSE images using the Tescan VEGA3 SEM at Vanderbilt University. We will also collect trace element profiles using line scans with consistently spaced laser pits from the ThermoFisher iCAP Qc quadrupole ICPMS at Vanderbilt University.

Hierarchical Clustering and Machine Learning in Python:

The skills to develop code for this project will be gained from taking DataCamp online courses and also by utilizing the abundant wealth of information on the internet. The hierarchical clustering algorithm will effectively calculate the Euclidean distance between all the points and store them in a proximity matrix. Points with the smallest distances separating them will be merged into clusters until only one single cluster is left. A dendrogram is then created and the optimum number of clusters is determined visually by intersecting the tallest vertical line in the dendrogram.

The machine learning algorithm will be trained to classify minerals into these pre-defined clusters based on major element data from an EDS map. In order to test the efficacy of this method, we will first test it on a large thin section of bulk tuff from the vitrophyre. If we can estimate the bulk eruptive material here, we can apply the methodology to a single stratigraphic section. If successful, we can apply this to a stratigraphic horizon in another part of the caldera and see if we observe similar results.

Spatial Statistics in ArcGIS:

Proficiency in ArcGIS was gained from an Advanced Spatial Analysis course taken in 2018 at Vanderbilt University. ArcGIS skills will be refined by participating in an ESRI MOOC titled Spatial Data Science: The New Frontier in Analytics which runs from February 12-April 9, 2020. The Spatial Analysis Research Laboratory (SARL) at Vanderbilt University will also provide expertise in GIS if any roadblocks may arise.

Ar-Ar Dating:

Ar-Ar sanidine ages will be obtained from the New Mexico Geochronology Research Laboratory (NMGRL) at New Mexico Tech. Sanidines will be separated, irradiated, and prepared for analysis using a Thermo Scientific Helix Multicollector. Funding for Ar-Ar dating will hopefully be provided by the Awards for Geochronology Student Research 2 (AGeS2) grant which will provide support for one year in 2020.

Timeline

Year 1 (2019-2020):

Year 2 (2020-2021):

Year 3 (2021-2022):

Broader Impacts

Our community has developed novel methods to assess the records preserved within individual crystals in order to understand the evolution of magmatic systems. With the expansion of open source science, we have capitalized on collaboration and creating workflows that are accessible to all. This proposed project will develop python scripts that will automate statistical analyses and tedious point counting to further the open science revolution!

Developing open-source science with python with hopes of uploading to the ENKI server for distribution.

Mentoring undergraduates! Female Power!!

References
