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Table of contents

Pı	eface	3							
1 Introduction									
2 Summary									
3	geochemistry	6							
	3.1 X-ray diffraction (XRD)	. 6							
	3.2 Scanning electron microscopy and energy dispersive X-ray spectroscopy								
	(SEM-EDS)	. 6							
	3.3 Portable X-ray fluorescence (pXRF)	. 6							
	3.4 Methodology	. 18							
	3.5 Results	. 19							
Re	ferences	20							

Preface

This is a Quarto book.

To learn more about Quarto books visit $\label{eq:https://quarto.org/docs/books.}$

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[1] 2

1 Introduction

This is a book created from markdown and executable code.

See Knuth (1984) for additional discussion of literate programming.

1 + 1

[1] 2

2 Summary

In summary, this book has no content whatsoever.

1 + 1

[1] 2

3 geochemistry

3.1 X-ray diffraction (XRD)

Some text. Some other text. A bit more of text.

3.2 Scanning electron microscopy and energy dispersive X-ray spectroscopy (SEM-EDS)

3.3 Portable X-ray fluorescence (pXRF)

Call:

PCA(X = df_scaled_alg, graph = FALSE)

Eigenvalues

Dim.1 Dim.2 Dim.3 Dim.4 Dim.5 Dim.6 Dim.7

Variance 2.419 1.842 1.009 0.800 0.663 0.151 0.116

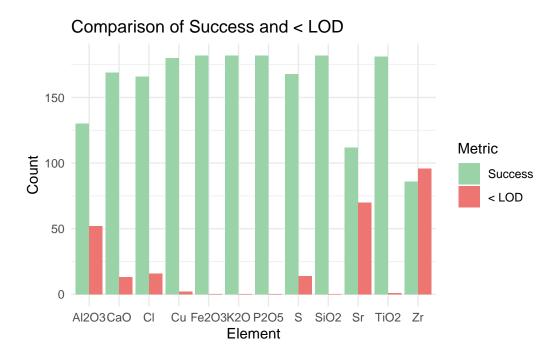


Figure 3.1: A barplot.

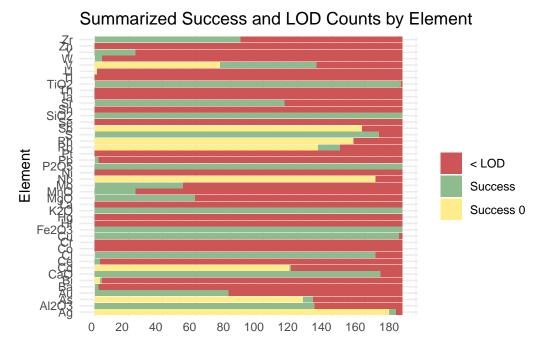


Figure 3.2: A barplot.

% of var. 34.558 26.319 14.408 11.424 9.477 2.154 1.660 Cumulative % of var. 34.558 60.877 75.284 86.708 96.186 98.340 100.000

Individuals (the 10 first)

Dist Dim.2 ctr Dim.3 Dim.1 ctr cos2 cos2 ctr LJW | 5.517 | 2.737 4.995 0.246 | -1.311 1.504 0.056 | 2.627 11.035 LJW.1 | 3.781 | 3.129 6.527 0.685 | -1.509 1.995 0.159 | -0.957 1.465 LJW.2 | 5.036 | 2.464 4.047 0.239 | 2.696 6.365 0.287 | 2.566 10.531 LJW.3 | 1.616 | 0.869 0.503 0.289 | -1.289 1.455 0.636 | -0.092 0.013 LJW.4 | 6.734 | 3.205 6.849 0.227 | 4.295 16.149 0.407 | 3.622 20.978 LJW.5 | 3.092 | -2.104 2.952 0.463 | 2.084 3.804 0.455 | -0.332 0.176 LJW.6 | 2.173 | 1.088 0.789 0.251 | -1.751 2.683 0.649 | -0.387 0.240 LJW.7 | 1.942 | 1.247 1.036 0.412 | -1.282 1.440 0.436 | -0.305 0.148 LJW.8 | 3.021 | 2.064 2.840 0.467 | -1.639 2.352 0.294 | -0.995 1.582 LJW.9 | 6.299 | 4.211 11.824 0.447 | 3.056 8.174 0.235 | -1.359 2.953

LJW 0.227 |

cos2

LJW.1 0.064 |

LJW.2 0.260 |

LJW.3 0.003 |

LJW.4 0.289 |

LJW.5 0.012 |

LJW.6 0.032 |

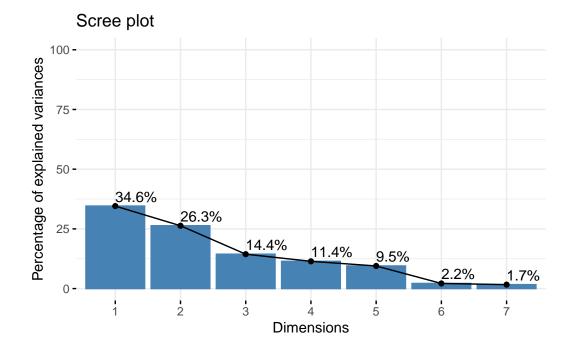
LJW.7 0.025 |

LJW.8 0.108 |

LJW.9 0.047 |

Variables

Dim.1 Dim.2 Dim.3 ctr cos2 ctr cos2 ctr cos2 0.805 26.793 0.490 13.038 P205 0.648 | 0.240 | -0.148 2.184 0.022 | S 0.659 17.966 0.435 | 0.491 13.067 0.241 | -0.154 2.358 0.024 | Cl0.360 5.358 0.130 | 0.474 12.190 0.225 | 0.611 37.034 0.374 | 0.445 | -0.675 24.757 K20 0.667 18.400 0.456 | -0.141 0.020 | CaO0.081 0.270 0.007 | 0.596 19.287 0.355 | -0.379 14.249 0.144 | 0.534 | -0.530 15.270 Ti02 0.730 22.056 0.281 | -0.223 0.050 I Fe203 | 0.471 9.157 0.222 | -0.210 2.391 0.044 | 0.613 37.279



Call:

PCA(X = df_scaled_nl, graph = FALSE)

Eigenvalues

	Dim.1	Dim.2	Dim.3	Dim.4	Dim.5	Dim.6
Variance	2.933	1.713	0.672	0.354	0.249	0.079
% of var.	48.885	28.546	11.197	5.901	4.157	1.314
Cumulative % of var.	48.885	77.431	88.628	94.529	98.686	100.000

Individuals (the 10 first)

	Dist Dim.1	ctr	cos2 Dim.2	ctr	cos2 Dim.3	ctr
CPT	1.744 -1.714	1.411	0.967 -0.066	0.004	0.001 -0.142	0.042
CPT.1	1.797 -1.356	0.884	0.570 0.402	0.133	0.050 1.046	2.295
CPT.2	1.426 -1.330	0.850	0.870 -0.353	0.103	0.061 -0.060	0.008
CPT.3	1.825 -1.803	1.561	0.975 -0.070	0.004	0.001 -0.084	0.015
CPT.4	1.815 -1.703	1.393	0.881 -0.332	0.090	0.033 -0.360	0.272
CPT.5	1.224 -0.702	0.237	0.329 -0.651	0.348	0.282 -0.659	0.911
CPT.6	1.542 -1.503	1.085	0.951 -0.100	0.008	0.004 -0.197	0.081
CPT.7	1.883 -1.787	1.534	0.901 -0.216	0.038	0.013 -0.432	0.391
CPT.8	1.592 -1.544	1.145	0.941 0.056	0.003	0.001 0.099	0.020
CPT.9	1.712 -1.523	1.114	0.792 0.441	0.160	0.066 0.475	0.474

cos2

CPT 0.007 |

CPT.1 0.339 |

CPT.2 0.002 |

CPT.3 0.002 |

CPT.4 0.039 |

CPT.5 0.290 |

CPT.6 0.016 |

CPT.7 0.053 |

CPT.8 0.004 |

CPT.9 0.077 |

Variables

Dim.1 ctr cos2 Dim.2 ctr cos2 Dim.3 ctr cos2

P205 | 0.668 15.194 0.446 | 0.591 20.375 0.349 | -0.119 2.094 0.014 |

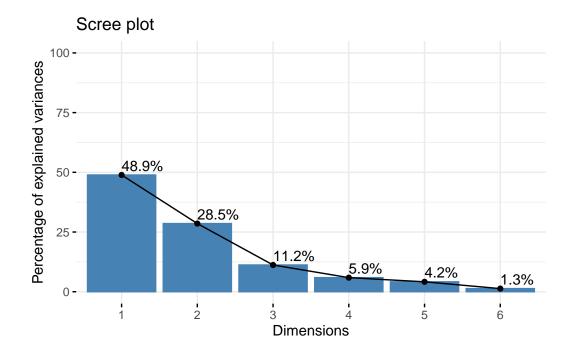
K20 | 0.836 23.802 0.698 | -0.463 12.504 0.214 | -0.073 0.791 0.005 |

Ti02 | 0.886 26.789 0.786 | -0.370 7.986 0.137 | -0.053 0.411 0.003 |

Fe203 | 0.844 24.304 0.713 | -0.237 3.278 0.056 | 0.110 1.797 0.012 |

Cu | 0.411 5.761 0.169 | 0.632 23.299 0.399 | 0.637 60.406 0.406 |

S | 0.349 4.150 0.122 | 0.747 32.557 0.558 | -0.481 34.501 0.232 |



Call:
PCA(X = df_scaled_comp, graph = FALSE)

Eigenvalues

	Dim.1	Dim.2	Dim.3	Dim.4	Dim.5	Dim.6
Variance	2.729	1.753	0.712	0.456	0.247	0.103
% of var.	45.477	29.222	11.871	7.600	4.121	1.709
Cumulative % of var.	45.477	74.699	86.570	94.170	98.291	100.000

Individuals (the 10 first)

Dist Dim.1 ctr cos2 Dim.2 ctr cos2 Dim.3 ctr

```
CPT | 1.872 | -1.795 | 1.639 | 0.919 | 0.369 | 0.108 | 0.039 | -0.111 | 0.024 | CPT.1 | 1.907 | -1.389 | 0.982 | 0.531 | 0.687 | 0.374 | 0.130 | 0.936 | 1.710 | CPT.2 | 1.356 | -1.275 | 0.827 | 0.884 | -0.107 | 0.009 | 0.006 | 0.041 | 0.003 | CPT.3 | 2.000 | -1.935 | 1.907 | 0.936 | 0.426 | 0.144 | 0.045 | -0.116 | 0.026 | CPT.4 | 1.850 | -1.773 | 1.600 | 0.918 | 0.089 | 0.006 | 0.002 | -0.328 | 0.209 | CPT.5 | 1.308 | -0.359 | 0.066 | 0.075 | -0.852 | 0.574 | 0.424 | -0.776 | 1.174 | CPT.6 | 1.570 | -1.502 | 1.149 | 0.915 | 0.237 | 0.045 | 0.023 | -0.150 | 0.044 | CPT.7 | 1.991 | -1.882 | 1.802 | 0.894 | 0.248 | 0.049 | 0.016 | -0.399 | 0.310 | CPT.8 | 1.674 | -1.552 | 1.225 | 0.860 | 0.353 | 0.099 | 0.044 | 0.044 | 0.004 | CPT.9 | 1.986 | -1.618 | 1.333 | 0.664 | 0.979 | 0.759 | 0.243 | 0.475 | 0.440 | cos2
```

CPT 0.003 |

CPT.1 0.241 |

CPT.2 0.001 |

CPT.3 0.003 |

CPT.4 0.031 |

CPT.5 0.352 |

CPT.6 0.009 |

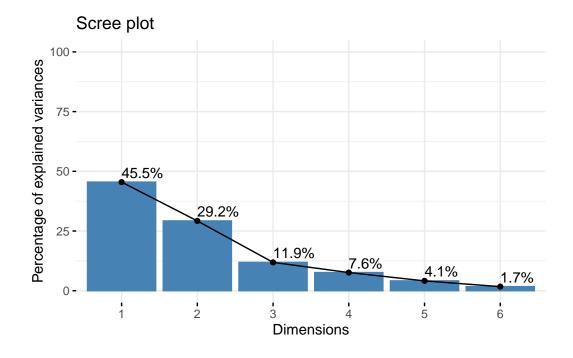
CPT.7 0.040 |

CPT.8 0.001 |

CPT.9 0.057 |

Variables

Dim.1 ctr cos2 Dim.2 cos2 Dim.3 cos2 ctr P205 0.663 16.108 0.440 | 0.620 21.955 0.385 | -0.190 5.080 0.036 | 0.507 | -0.630 22.626 K20 0.712 18.598 0.397 | -0.108 1.653 0.012 | 0.800 23.482 0.641 | -0.491 13.763 0.241 | -0.048 0.321 0.002 | Fe203 | 0.751 20.672 0.564 | -0.247 3.491 0.061 | 0.295 12.221 0.087 | Cu 0.463 7.866 0.215 | 0.576 18.943 0.332 | 0.617 53.383 0.380 | S 0.602 13.273 0.362 | 0.581 19.222 0.337 | -0.441 27.342



Call:
PCA(X = df_scaled_comp2, graph = FALSE)

Eigenvalues

Dim.1 Dim.2 Dim.3 Dim.4 Dim.5 Dim.6

Variance 3.074 1.638 0.687 0.329 0.237 0.035

% of var. 51.229 27.306 11.444 5.478 3.956 0.588

Cumulative % of var. 51.229 78.534 89.978 95.456 99.412 100.000

Individuals (the 10 first)

Dist Dim.1 ctr cos2 Dim.2 ctr cos2 Dim.3 ctr

LJW | 3.502 | 1.688 0.997 0.232 | -0.419 0.115 0.014 | 0.306 0.147

LJW.1 | 1.452 | 1.128 0.445 0.604 | 0.594 0.232 0.168 | -0.644 0.649

LJW.2 | 2.521 | 0.866 0.262 0.118 | 1.892 2.350 0.563 | -1.389 3.020

LJW.3 | 0.560 | 0.044 0.001 0.006 | -0.390 0.100 0.485 | -0.313 0.153

LJW.4 | 3.148 | 1.069 0.400 0.115 | 2.767 5.026 0.773 | -0.431 0.290

LJW.5 | 1.545 | -1.026 0.368 0.441 | 0.339 0.076 0.048 | -1.094 1.876

LJW.6 | 0.899 | 0.196 0.013 0.048 | -0.082 0.004 0.008 | -0.832 1.084

LJW.7 | 1.453 | -0.219 0.017 0.023 | -0.853 0.478 0.345 | 0.961 1.445

LJW.8 | 0.819 | 0.167 0.010 0.041 | -0.546 0.196 0.445 | 0.524 0.430

LJW.9 | 6.112 | 1.021 0.365 0.028 | 4.521 13.412 0.547 | 3.411 18.220

cos2

LJW 0.008 |

LJW.1 0.197 |

LJW.2 0.303 |

LJW.3 0.313 |

LJW.4 0.019 |

LJW.5 0.502 |

LJW.6 0.857 |

LJW.7 0.437 |

LJW.8 0.410 |

LJW.9 0.311 |

Variables

		Dim.1	ctr	cos2	Dim.2	ctr	cos2	Dim.3	ctr	cos2
P205	I	0.646	13.594	0.418	0.600	21.968	0.360	0.160	3.721	0.026
Ti02	1	0.921	27.594	0.848	-0.316	6.092	0.100	0.052	0.399	0.003
K20	1	0.889	25.723	0.791	-0.371	8.410	0.138	0.114	1.903	0.013
Fe203	I	0.886	25.565	0.786	-0.180	1.986	0.033	-0.015	0.035	0.000
Cu	I	0.449	6.564	0.202	0.555	18.818	0.308	-0.690	69.388	0.476
S	ı	0.172	0.960	0.029	0.837	42.727	0.700	0.411	24.554	0.169

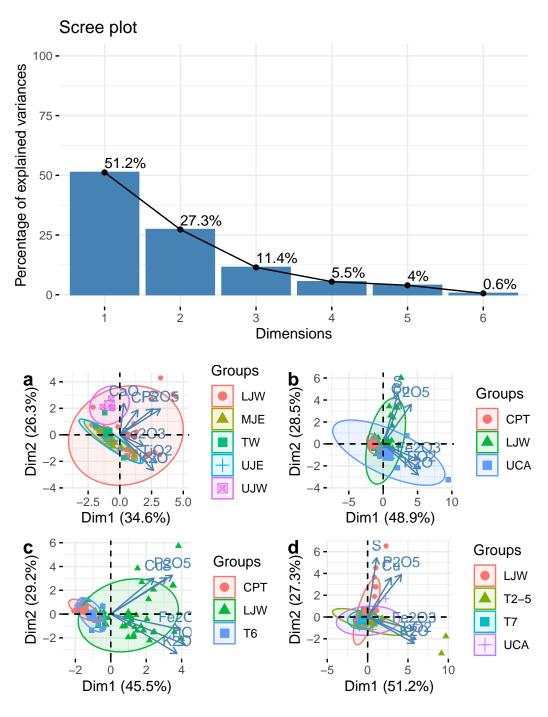


Figure 3.3: Several pcas

3.4 Methodology

For this experiment, a Bruker portable XRF Titan S1 was used in a laboratory benchtop setup, using battery power (up to 25% battery charge then replaced by a fully charged battery). A validation run was applied on two standard samples provided by Bruker and using the standard calibration. Samples were scanned for 180 seconds each (90 seconds for the first phase for major elements and 90 seconds for the second phase for minor elements), at least once, with several scans applied to samples that showed macroscopic variability (e.g., areas with different colours or translucency). The standard database from Bruker was used, with the Geochem application and Dual Mining method. ** chert samples were scanned, from several sources and chert types, both geological and archaeological. After scanning, the scanned face was measured (thickness and diameter), to guarantee a minimum thickness and size was followed for each sample, since other studies have shown thickness and size may impact the homogeneity of data collection and results (Newlander et al. 2015). All samples, including their thickness and diameter, can be found in table X. For geological samples, fresh, flat surfaces were scanned, avoiding altered faces or cortex; whenever necessary, the samples were prepared by breaking the nodules. The samples were chosen to represent all varieties of chert identified in the Algarve region, in the archaeological record of Vale Boi, but also from other regions such as Central Portugal and South Spain, to allow their comparison and test hypotheses made from macroscopic and petrographic data. For archaeological samples, artefacts were chosen from previously identified types (REF?), focusing on larger and flatter morphologies, with the least degree of surface alterations possible.

The analysis and result reporting protocol was established following previous studies, focusing on the accuracy of obtained data but also the transparency and reproducibility of the results (Newlander et al. 2015; Johnson et al. 2024). For further reproducibility and replicability, and working towards the goal of open science (Johnson et al. 2024; Marwick 2017), the obtained raw pXRF results can be found online on our online compendium (LINK).

3.5 Results

The pXRF measured several major and minor elements, of which a small amount returned values of 0 or were below the limit of detection (<LOD). These were uranium (U), thorium (Th), bismuth (Bi), thallium (Tl), mercury (Hg), platinum (Pt), tantalum (Ta), hafnium (Hf), lanthanum (La), antimony (Sb), Tin (Sn), cadmium (Cd), rhodium (Rh), niobium (Nb), selenium (Se), zinc (Zn), nickel (Ni), cobalt (Co) and chromium (Cr). They were removed from the analysis based on their nonexistence, although they can still be found in the raw pXRF results.

References

- Johnson, Kimberly, Colin P. Quinn, Nathan Goodale, and Richard Conrey. 2024. "Best Practices for Publishing pXRF Analyses." Advances in Archaeological Practice 12 (2): 156–62. https://doi.org/10.1017/aap.2024.6.
- Knuth, Donald E. 1984. "Literate Programming." Comput. J. 27 (2): 97–111. https://doi.org/10.1093/comjnl/27.2.97.
- Marwick, Ben. 2017. "Computational Reproducibility in Archaeological Research: Basic Principles and a Case Study of Their Implementation." Journal of Archaeological Method and Theory 24 (2): 424–50. https://doi.org/10.1007/s10816-015-9272-9.
- Newlander, Khori, Nathan Goodale, George T. Jones, and David G. Bailey. 2015. "Empirical Study of the Effect of Count Time on the Precision and Accuracy of pXRF Data." Journal of Archaeological Science: Reports 3 (September): 534–48. https://doi.org/10.1016/j.jasrep.2015.07.007.