

Assessing the Variability in Ore Hardness and Model Transferability for a Mineral Sands Deposit

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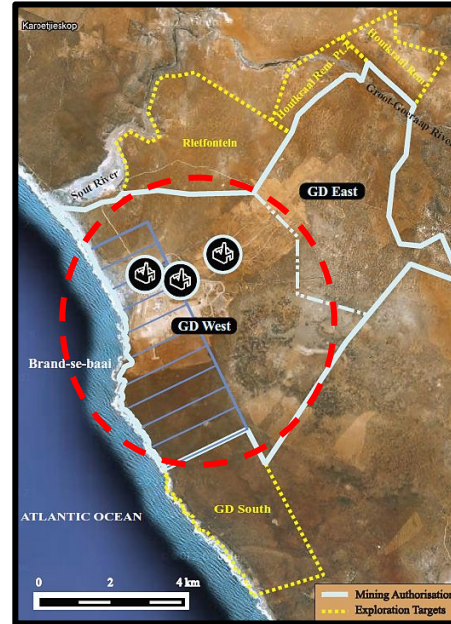
1. Centre for Minerals Research, University of Cape Town
2. Tronox Mineral Sands, South Africa

14-15 October 2025

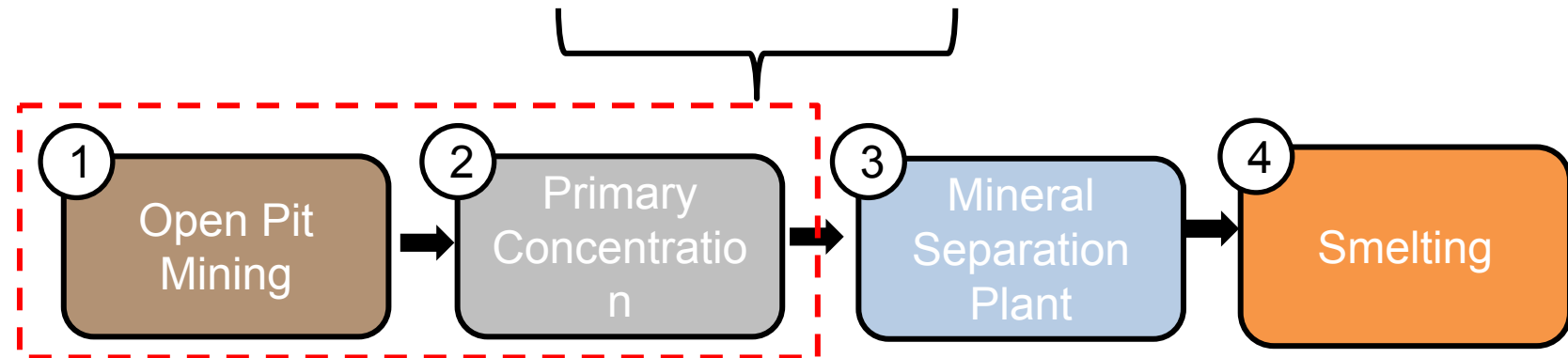
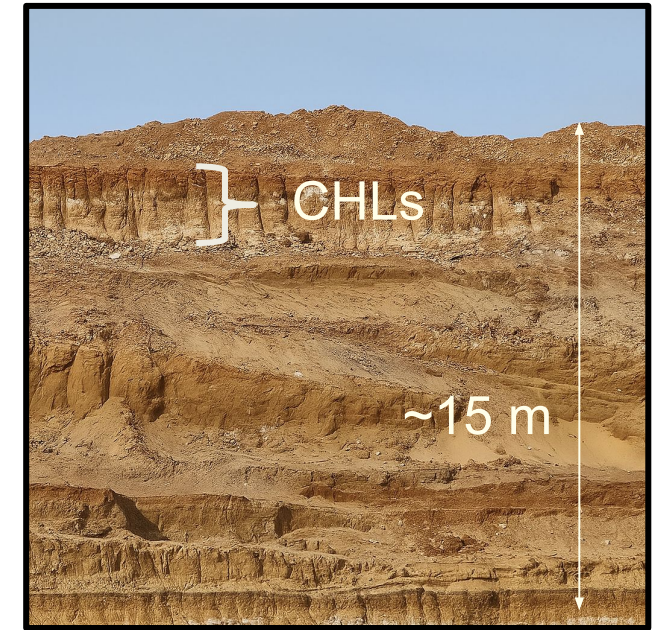


Introduction to mineral sands

- Location: Situated on South Africa's west coast
- Deposit hosts world-class heavy minerals (HMs) rutile (TiO_2), ilmenite (FeTiO_3) and zircon (ZrSiO_4)
- Unique geology: Formation of cemented hard layers ("CHLs") from normally loose sands

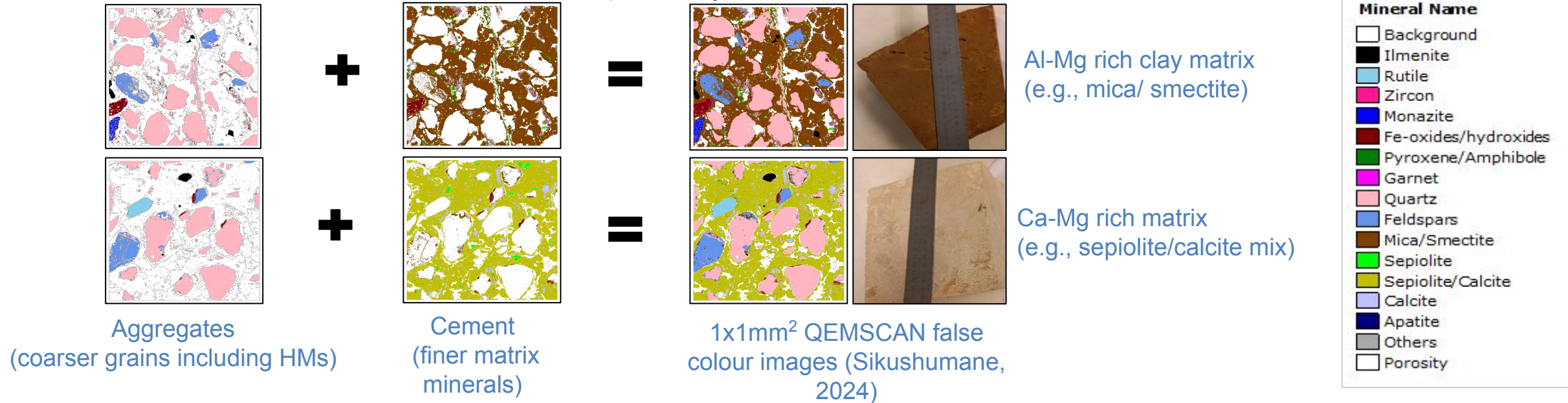


Philander & Rozendaal, 2015



Mineralogy of the CHLs

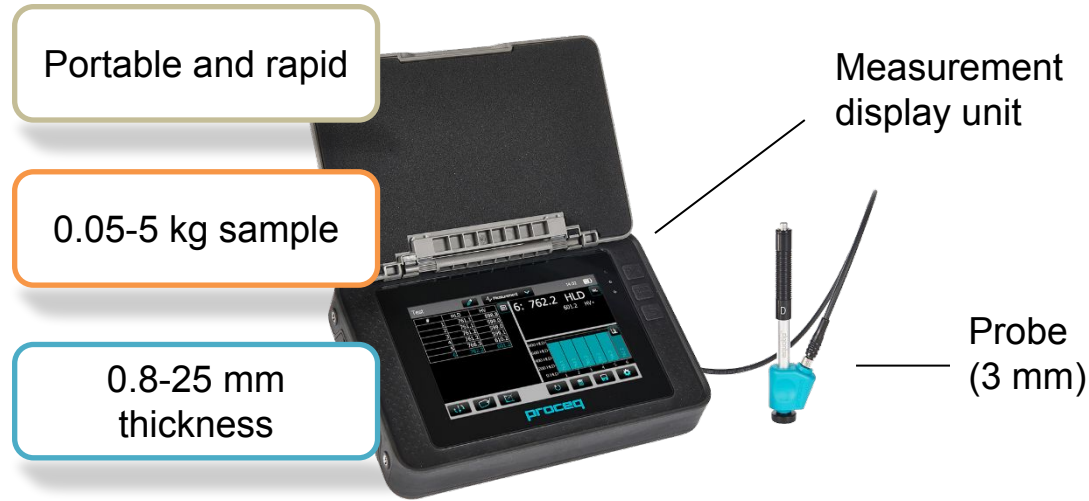
- Formed as a duricrust which precipitated minerals (e.g. sepiolite, calcite, mica/smectite) cemented the sand grains and valuable minerals
- Multiple periods of ground water activity - multiple CHL with varying mineralogy
- Theory: CHLs behave like concrete – matrix minerals bring strength and hardness to the CHL, along with other factors such as texture and porosity



- Variation in hardness resulted in the installation of a SAG mill in 2007
- Current geometallurgical models focus on grade and recovery, with no physical hardness proxy linked to processing behavior

Quantifying ore hardness for this study

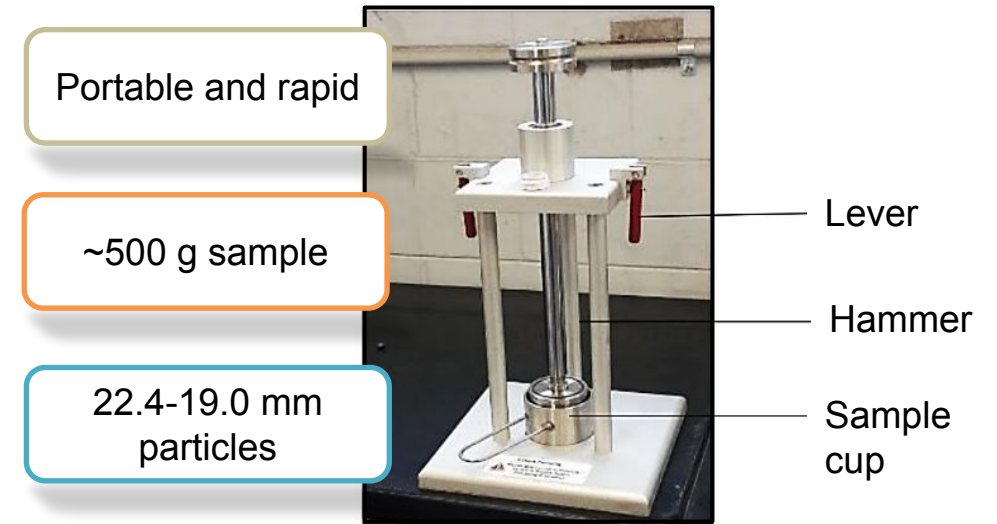
Leeb Hardness Tester



Proceq, 2017¹

- Proxy
- Non-destructive
- Measures the ratio of rebound to impact velocity
- HLS scale: 0 (soft) – 1000 (hard)

Hardness Index Tester (HIT)



Kojovic, 2016²

- Proxy
- Destructive
- Measures the relationship between impact energy and particle breakage
- A*b scale: <30 hard, >70 soft

1. Proceq, 2017. Equotip Portable Hardness Testing.

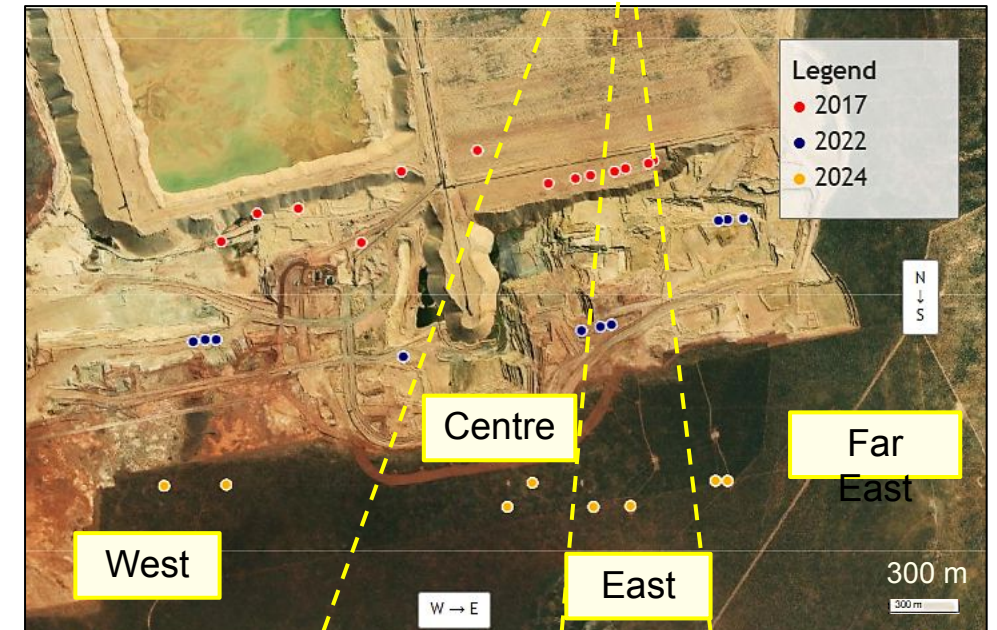
2. Kojovic, T. 2016. HIT-a Portable Field Device for Rapid Testing at Site. In 13th AusIMM Mill Operators Conference. Perth, West Australia. 9–16.

Study datasets: 2017, 2022, and 2024

Approach	2017 (39 samples)	2022* (43 samples)	2024 (24 samples)	Total
Type of sampling	Bulk	Selective	Bulk	-
Modelling HIT A*b				-
Bulk chemistry (XRF)	3	43	24	106
Mineralogy (XRD and QEMSCAN)		43	24	67
Leeb hardness (HLS)	19	43	24	86
Impact breakage (HIT A*b)	24	43	24	91
Mechanical strength (UCS ^a)			24	24
Mechanical strength (PLT)			24	24
Grindability (BWI)			24	24
	Not measured			
	Included in the analysis			
	Available, but not included in scope of presentation			

* 2022 dataset from Sikushumane (2024)

Sample distribution map showing arbitrary pit location divides (depth not represented)



Lateral variability (pit location)

Year	West (32)	Centre (19)	East (33)	Far East (22)
2017	15	3	18	3
2022	11	10	9	13
2024	6	6	6	6

Aim and key questions

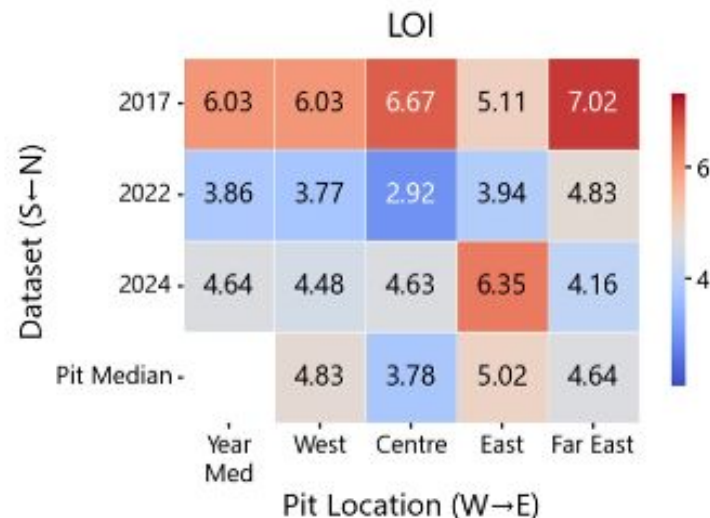
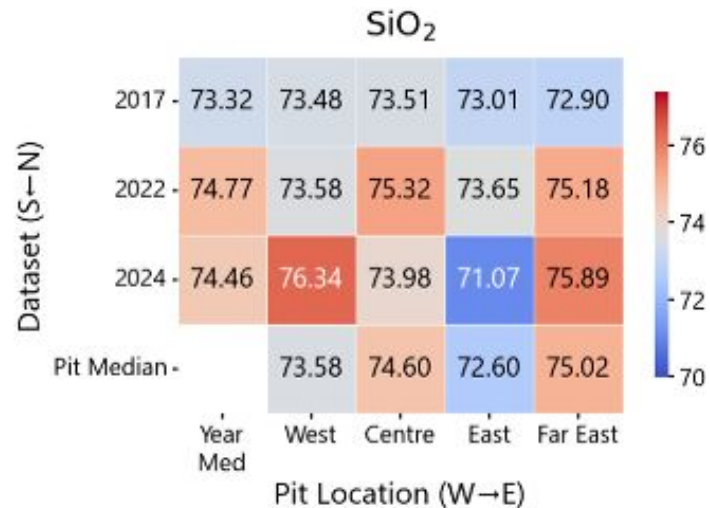
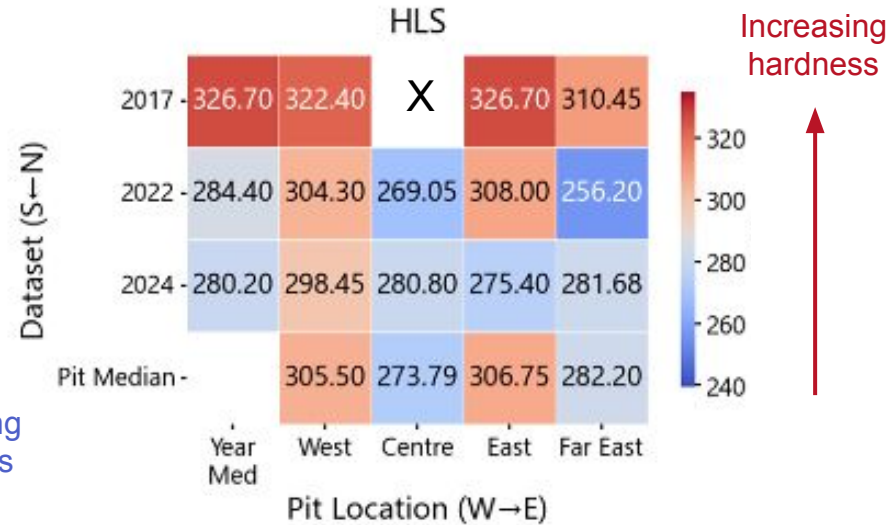
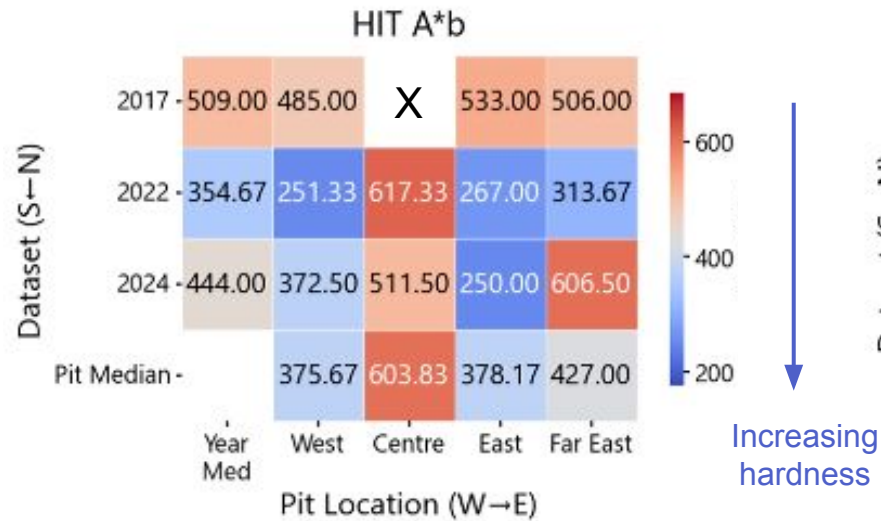
Aim:

To understand the variability in hardness with mineralogy and to test the transferability of the model developed by Sikushumane et al (submitted) across the different datasets

Key Questions:

1. How do chemistry and hardness relationships differ by sampling year?
2. Can a model trained on one year predict hardness in others?
3. What geological or sampling factors explain variability?

Variability across datasets and pit location



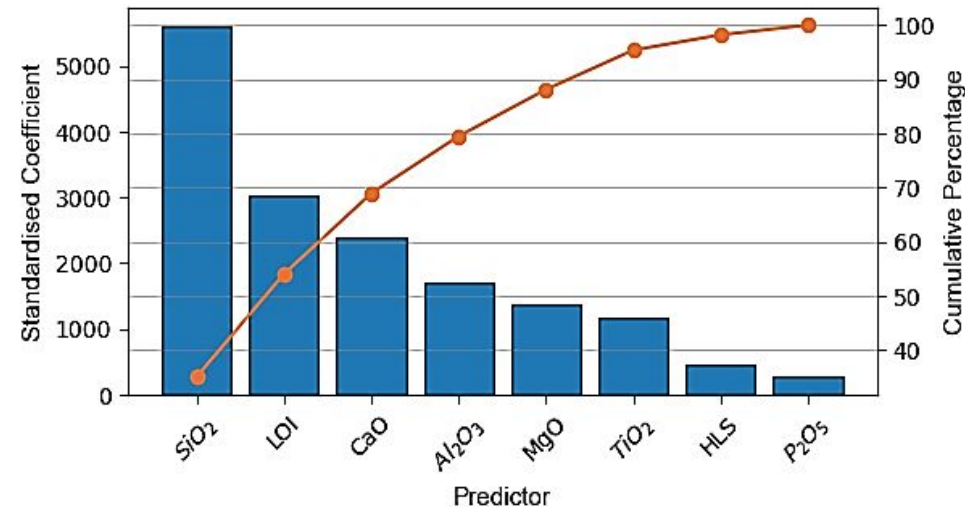
- Deposit is generally very soft.
- HIT A*b and HLS are more strongly related across pits than across years
- West and East pits are hardest; Centre pit consistently soft across years.
- Centre pit has high SiO₂ (granular quartz) and low LOI (less cementing material).

Modelling the relationship between chemistry, HLS and HIT A*b

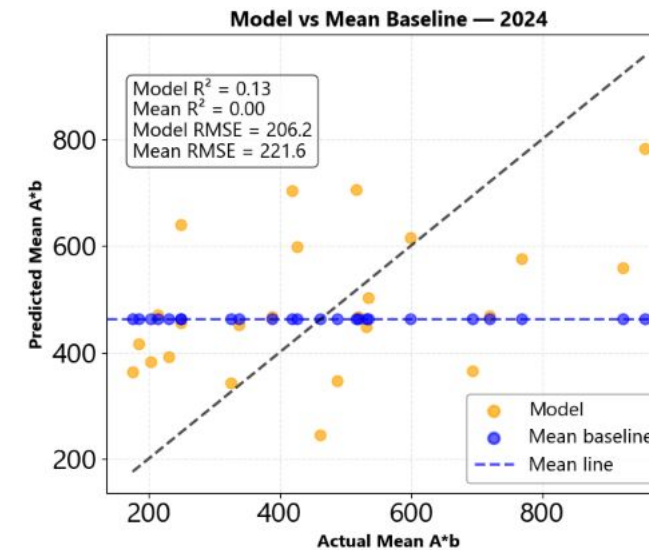
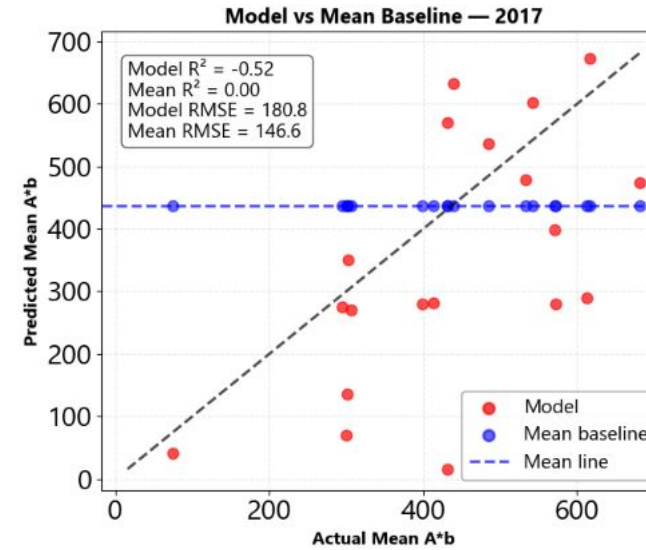
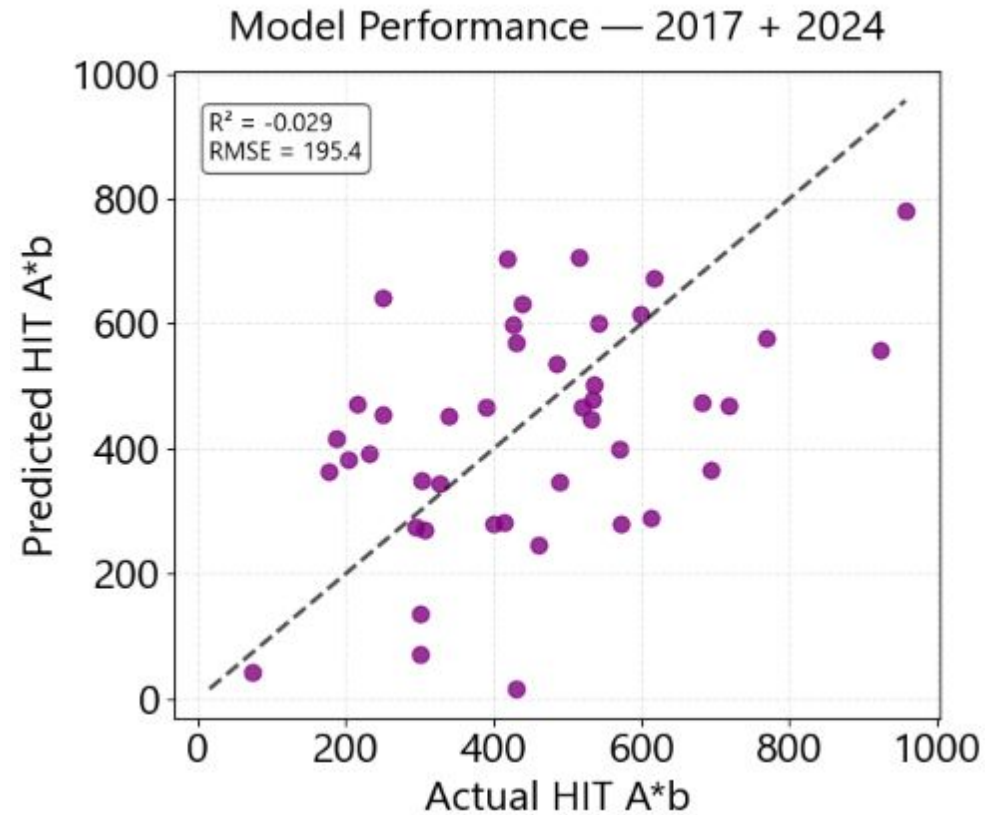
- Initial model developed on the 2022 dataset (43 samples)
- LOI reflects the loss of volatiles like water and CO₂ from clays and carbonates during XRF analysis

$$HIT A * b = 1.64 * 10^4 - 374TiO_2 - 524P_2O_5 - 132CaO - 150SiO_2 - 318Al_2O_3 - 221MgO - 165LOI - 2HLS$$

$R^2 = 0.80, RMSE = 114$ (root mean square error)

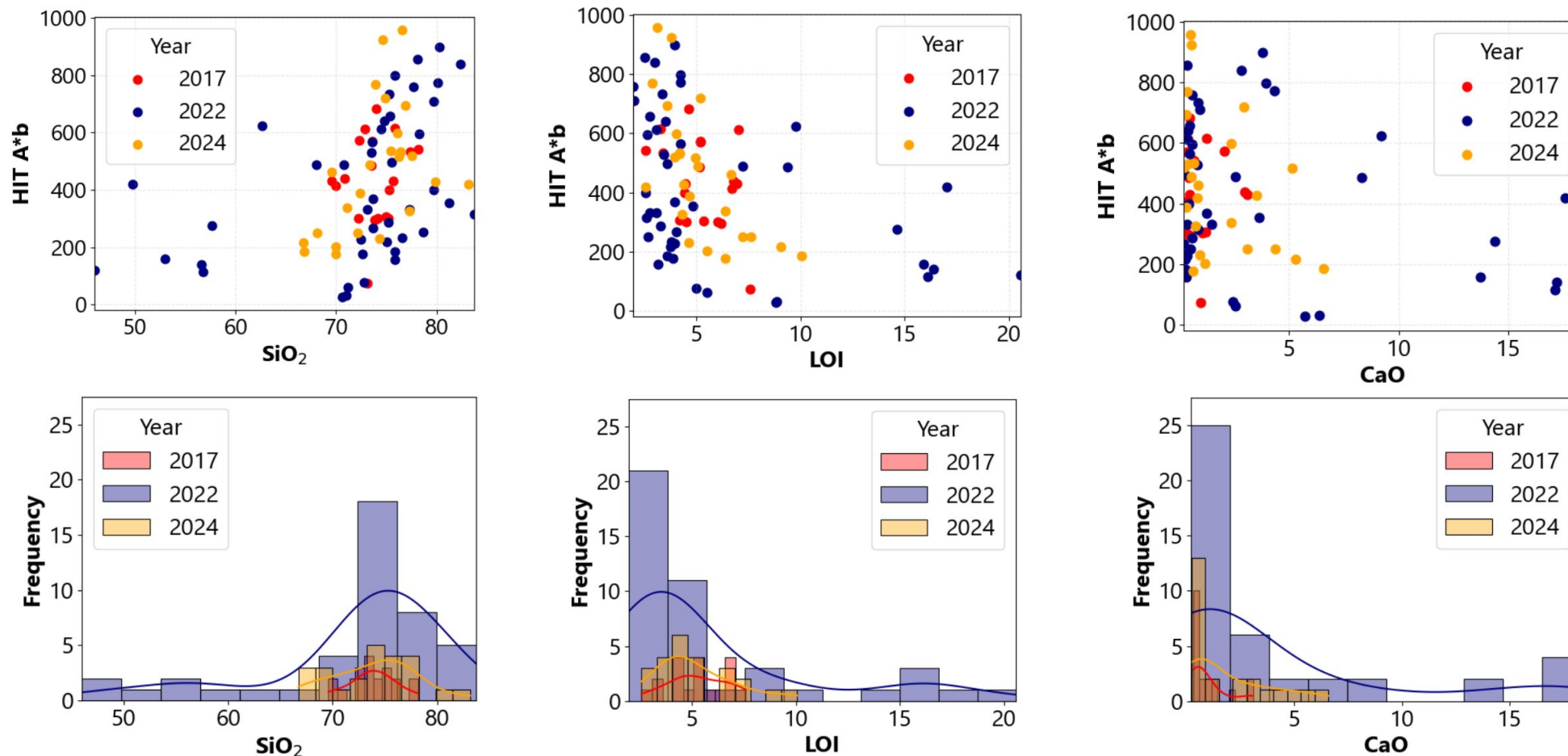


Model validation on 2017 and 2024 data

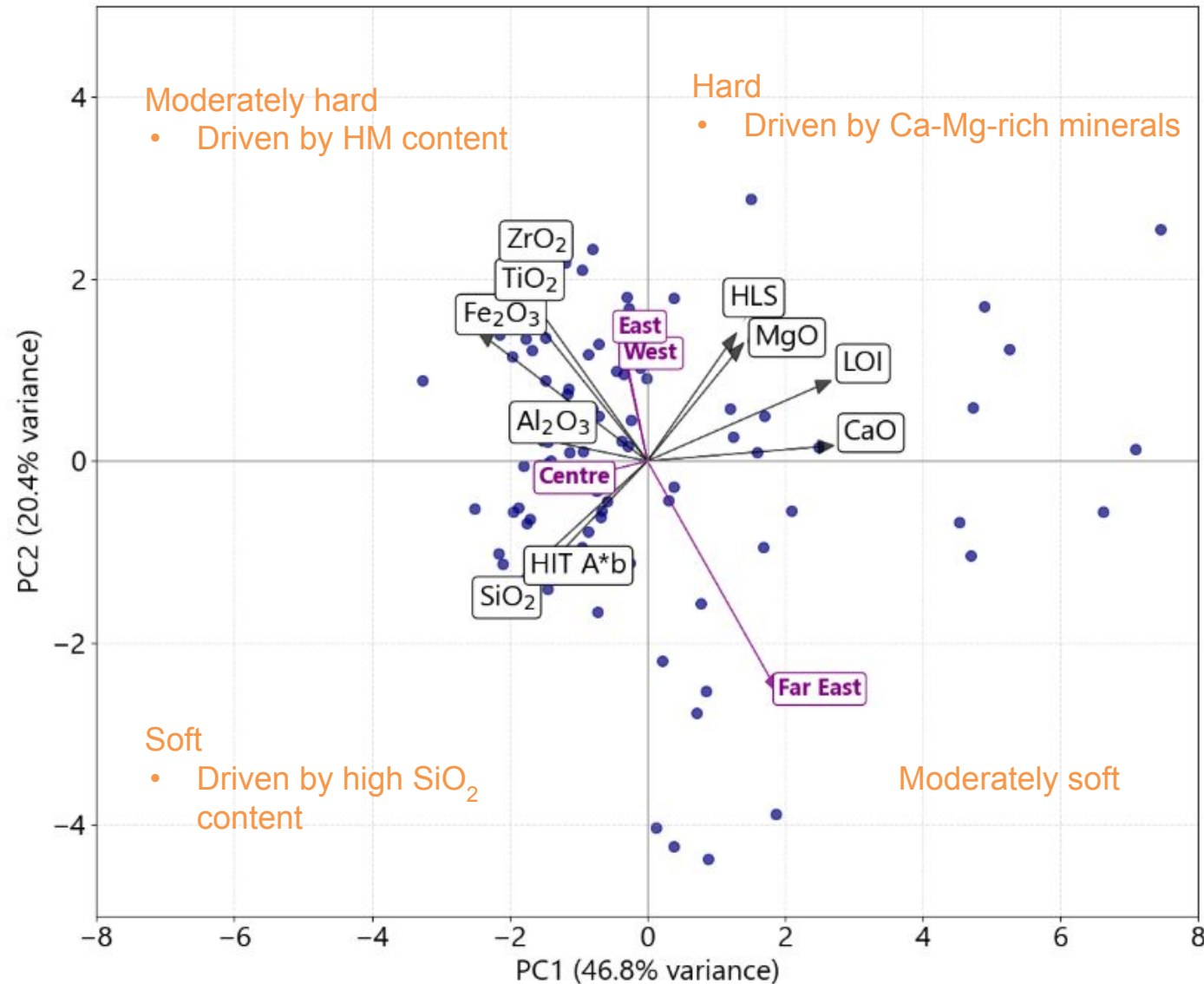


Understanding model performance: Features & Data Spread

- 2017 and 2024 data are chemically clustered, showing little of the variability needed for the model's predictions to hold

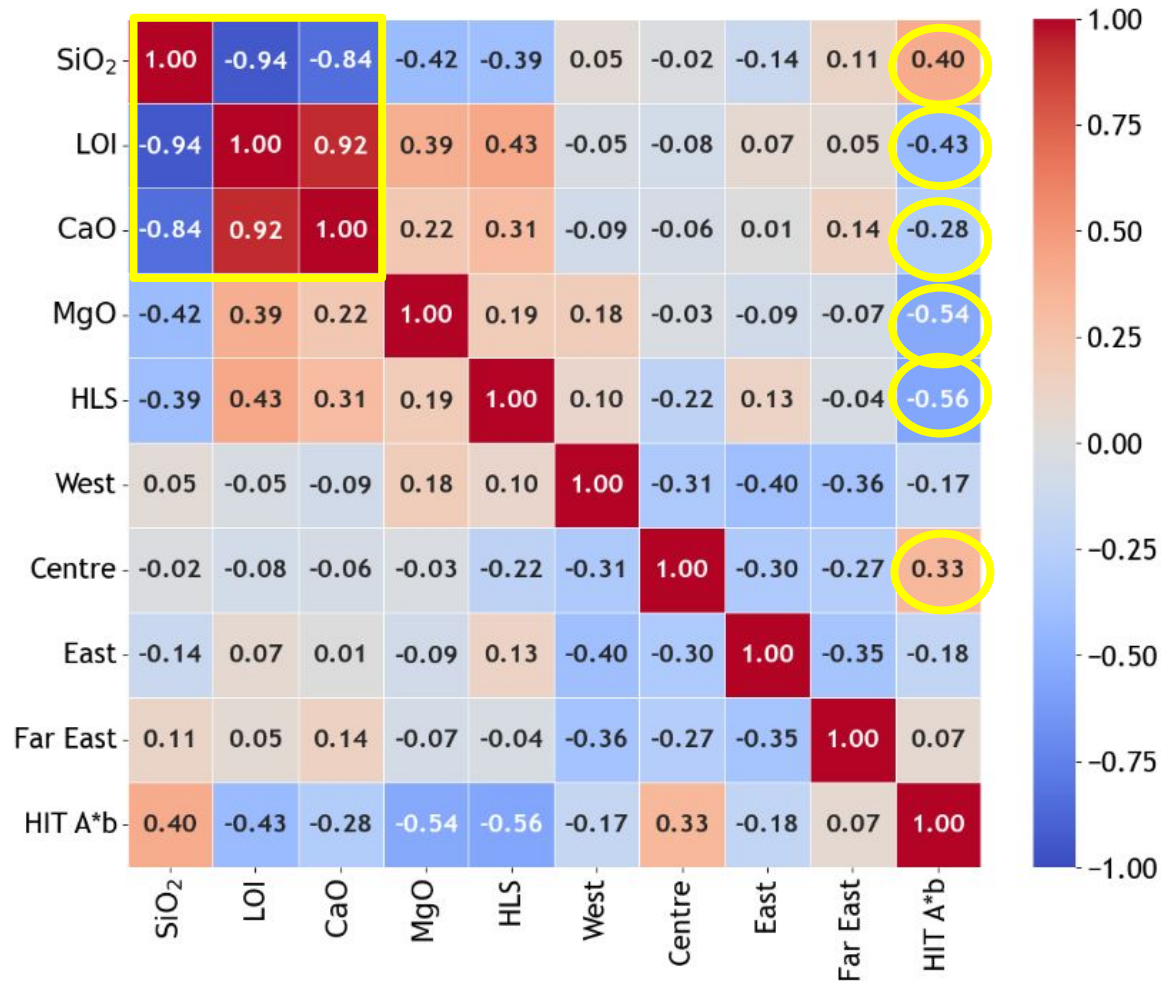


Principal component analysis (PCA)



- Reveals distinct geochemical domains
- PC1 (47%) separates "Grain-Dominant" (High SiO_2) from "Matrix-Dominant" (High CaO/LOI) ores
- PC2 (20%) separates soft Centre Pit material from hard Far East Pit material.

Moving towards improved models using combined dataset



- Strongest relationship to HIT A*b is HLS followed by MgO content
- Consideration of covariance: SiO₂, LOI and CaO

$|R| > 0.21$ statistically significant at 95% confidence interval for $n = 86$

Conclusions and next steps

- Hardness is systematically controlled by a "concrete-like" grain (SiO_2) vs. cement ($\text{CaO}/\text{LOI}/\text{MgO}$) relationship.
- The deposit contains distinct hard (East Pit) and soft (Centre Pit) domains, revealed by PCA.
- The developed model (Sikushumane et al. submitted) captured key chemical controls but showed limited transferability across the datasets.
- The study has also highlighted that consistent sampling is crucial for reliable classification and more accurate interpretation of hardness and chemistry trends.
- Further data analysis and model refinement are needed to address correlated variables and uneven data distribution, and to explore domain-specific.

Acknowledgements

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