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Automated sensory attribute consolidation:

Efficient integration with data analysis applied to craft gins and beers

13 September 2023

Sasev 2023

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Our proposal

Bigger net with bigger holes

- Instead of benchmarking products together, we got sensory descriptions from a wider array of professional sources
- Creating a sensible sensory map
- Investigating and testing automation and expansion of this concept



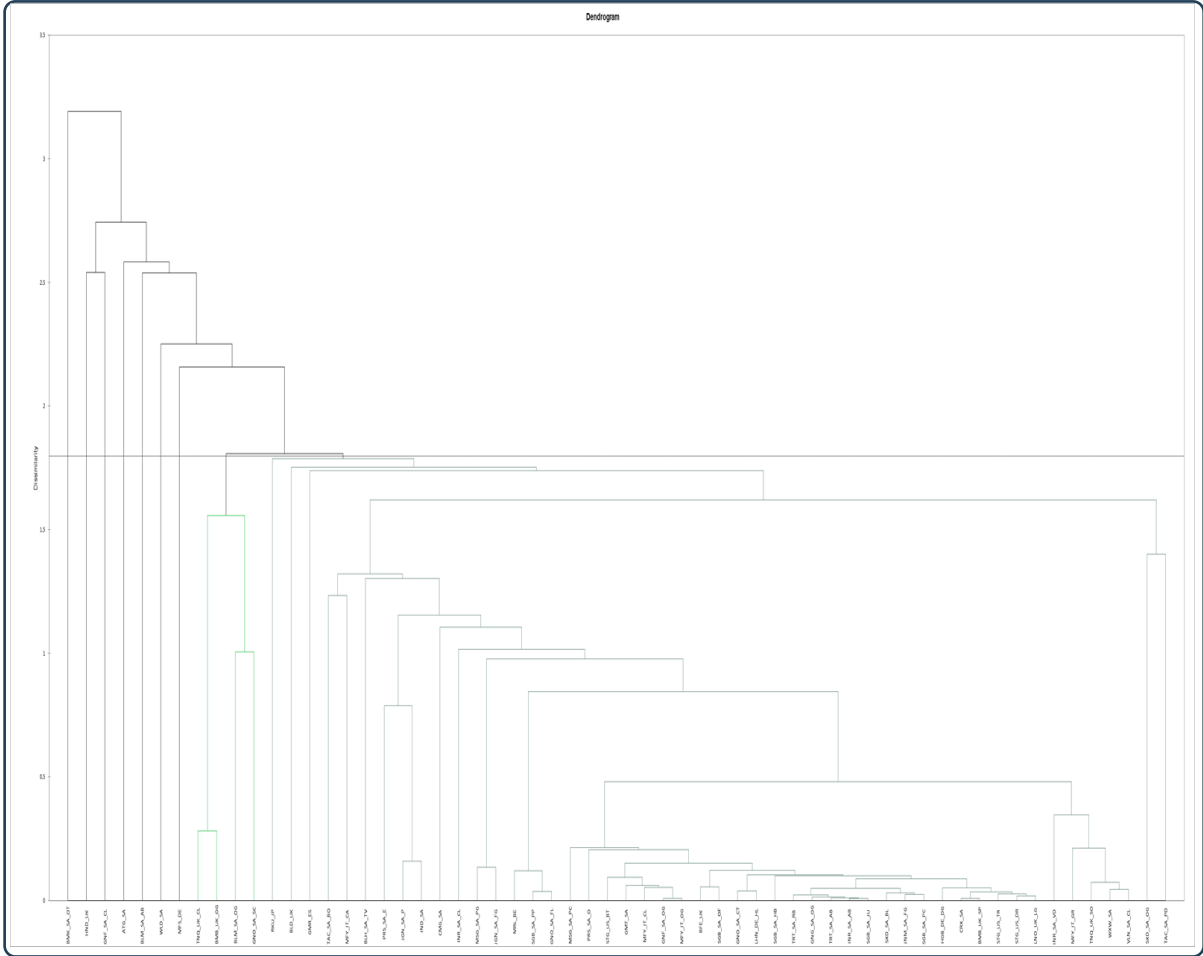
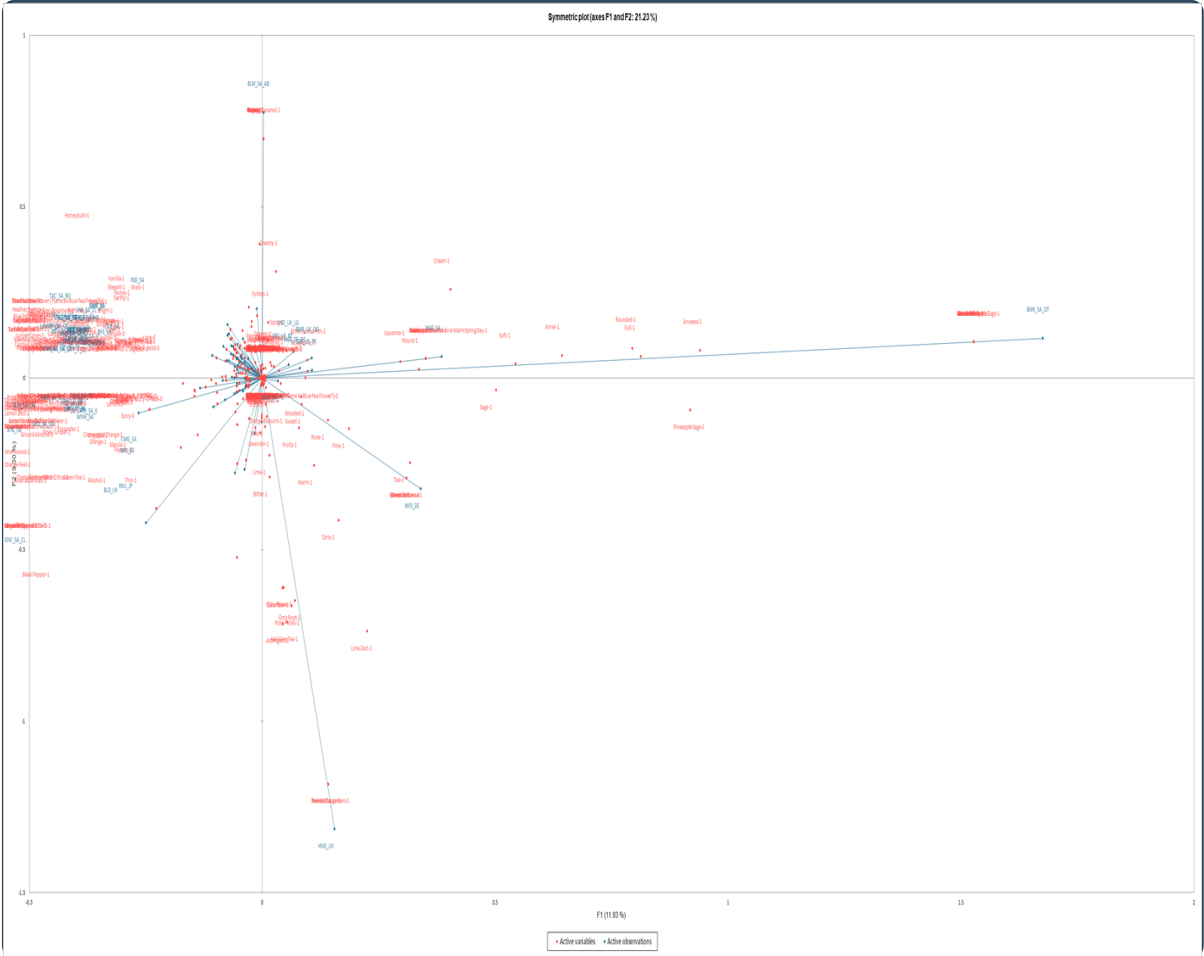
Maps, what are they good for?

- To show the representative value of each map, and to verify that distinctiveness and inter relationships between products where preserved

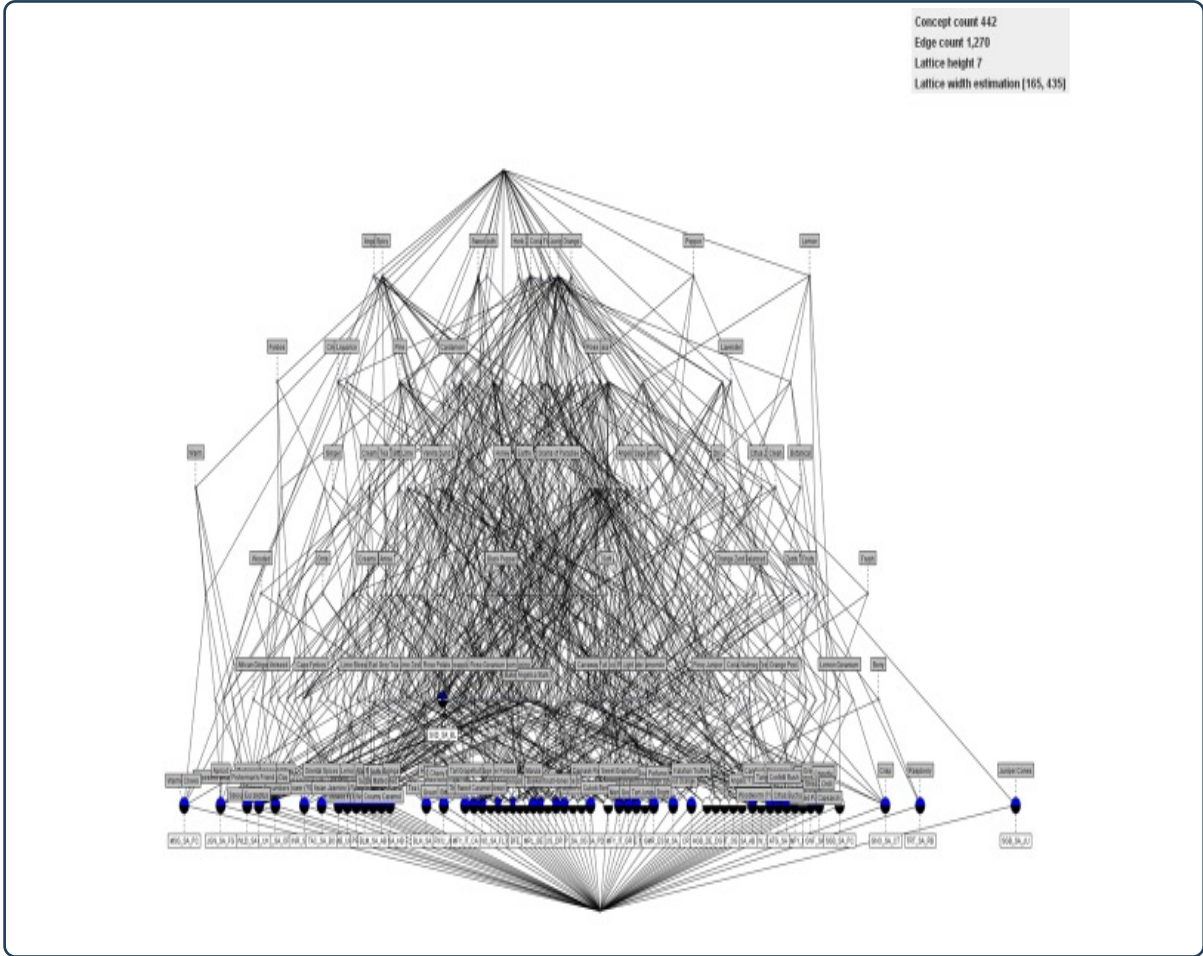
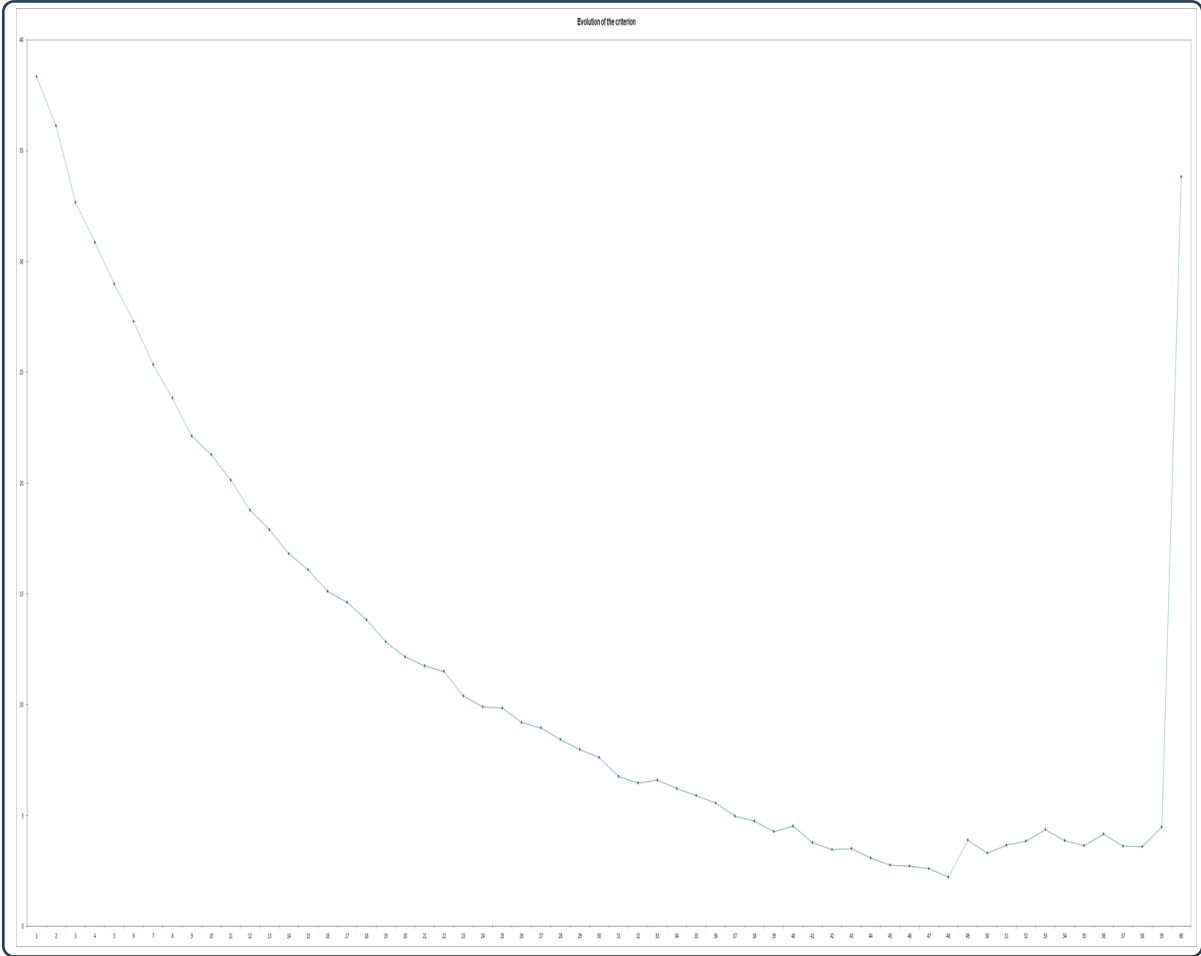
How was this verified?

- Spaces investigated with HAC, MCA, fKMeans, and Concept Lattices

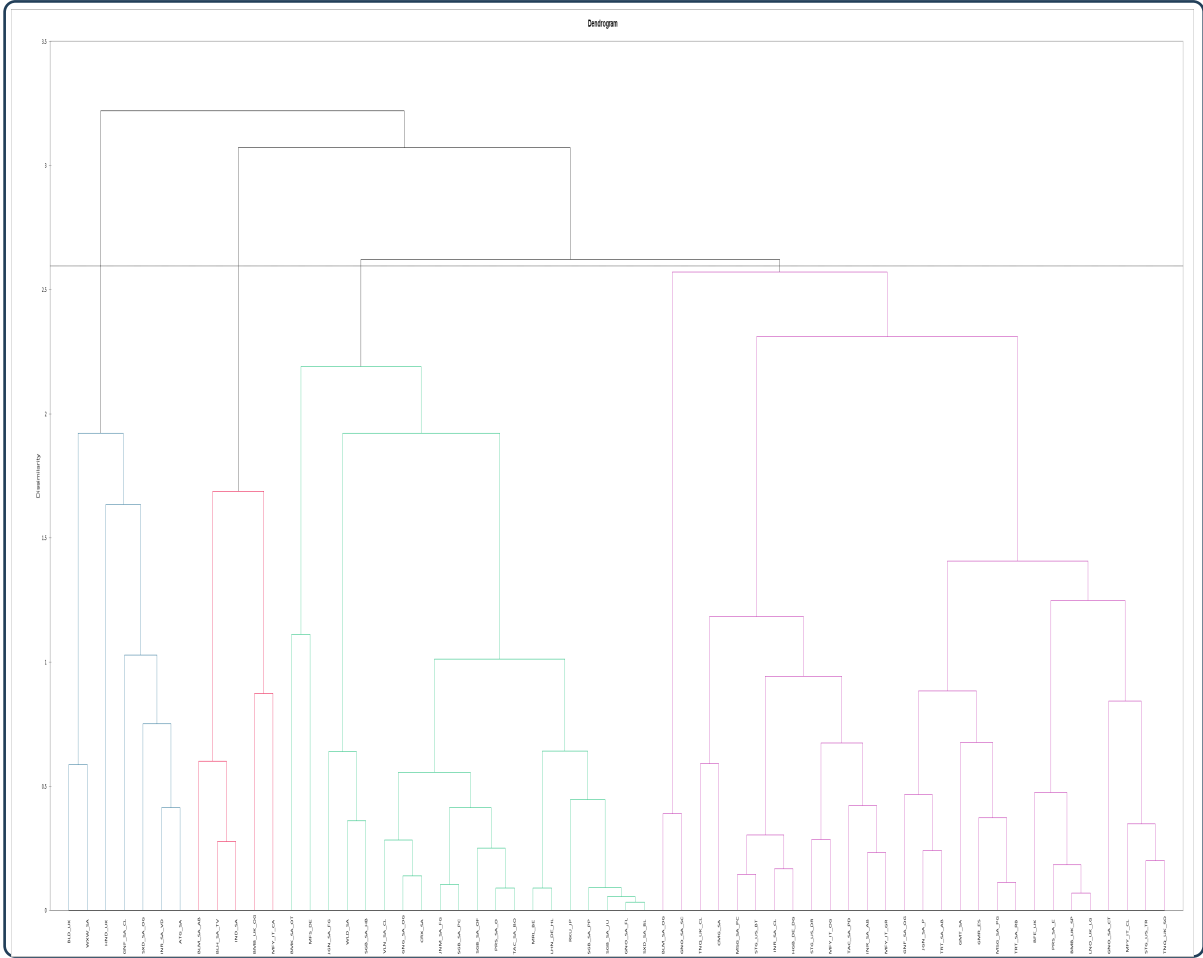
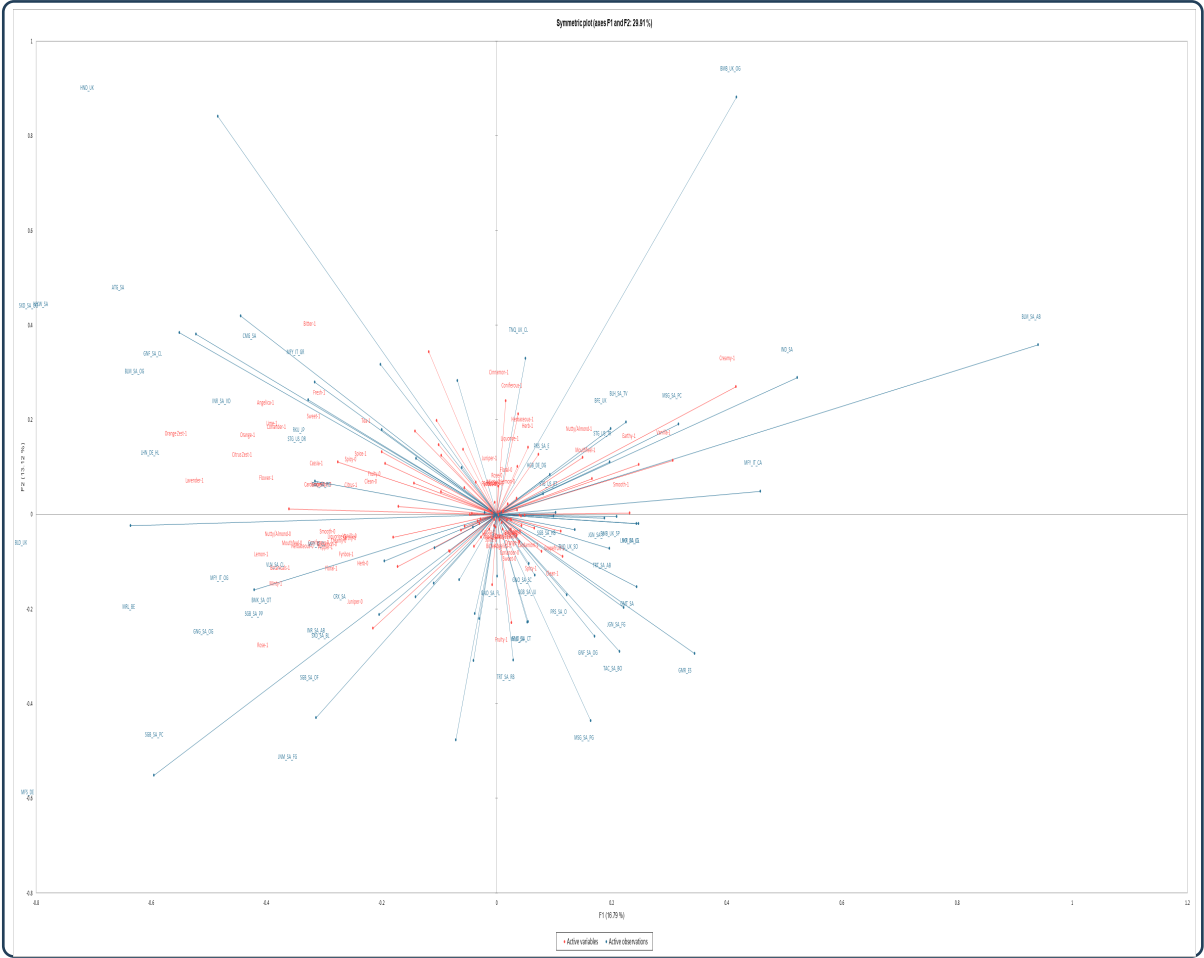
Application: Craft gin results from raw data analysis



Application: Craft gin results from raw data analysis



Application: Craft gin results from consolidated data analysis



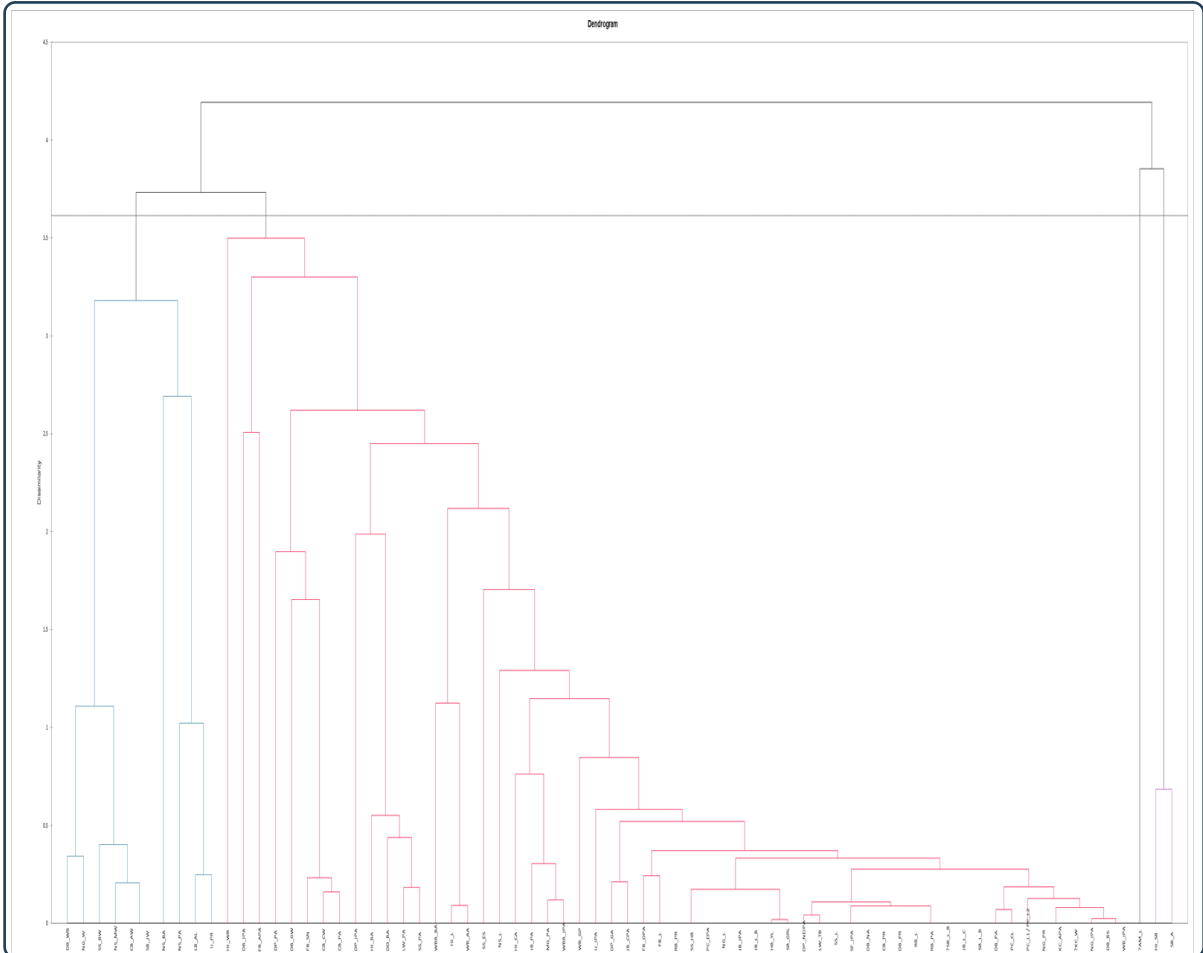
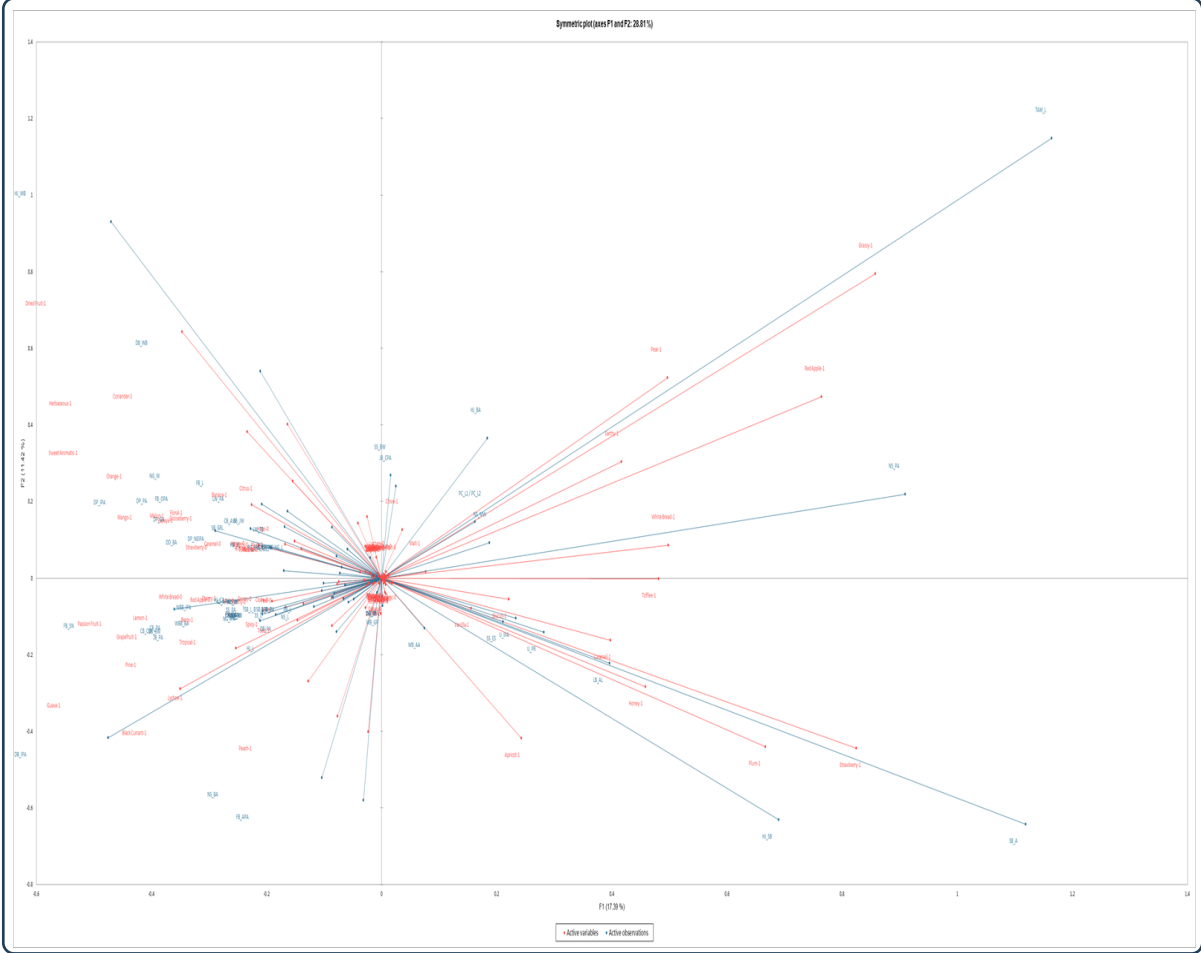
Application (Gin model comparison)

Comparison Metric?

To verify that our choices lead to comparable results with use RV coefficients to compare the different choices.

Datasets	Coverage	Dimensions	RV coefficients	Adjusted RV coefficients	p-value	Correlation
HRAW vs ARAW	100%	All	0.697	0.431	0.000	Low
HRAW vs HCON	100%	All	0.760	0.625	0.000	Moderate
ARAW vs AMAP	100%	All	0.786	0.656	0.000	Moderate
ARAW vs ACON	100%	All	0.822	0.727	0.000	High
HCON vs ACON	100%	All	0.446	0.246	0.000	Negligible

Application: Craft beer results from consolidated data analysis



Application (Beer model comparison)

Comparison Metric?

To verify that our choices lead to comparable results with use RV coefficients to compare the different choices.

Datasets	Coverage	Dimensions	RV coefficients	Adjusted RV coefficients	p-value	Correlation
HRAW vs ARAW	100%	All	0.411	0.229	0.000	Negligible
HRAW vs HCON	100%	All	0.527	0.373	0.000	Low
ARAW vs AMAP	100%	All	0.538	0.426	0.000	Low
ARAW vs ACON	100%	All	0.569	0.466	0.000	Low
HCON vs ACON	100%	All	0.841	0.801	0.000	High

Application (gin and beer)

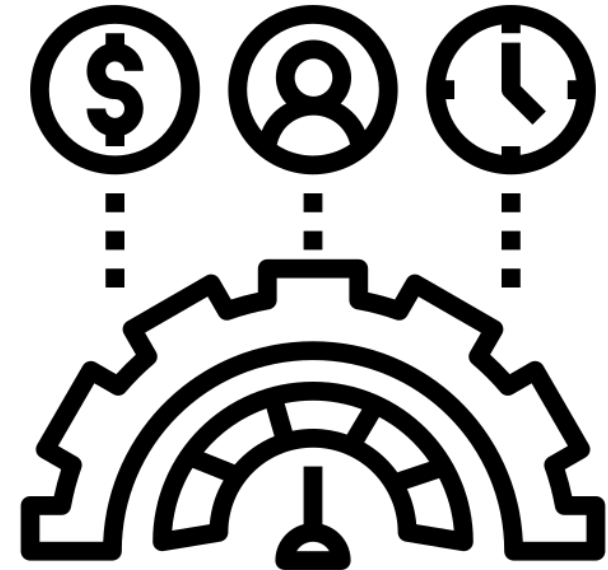
Does it work?

Yes (for a given value of work)

But there are shortcomings...

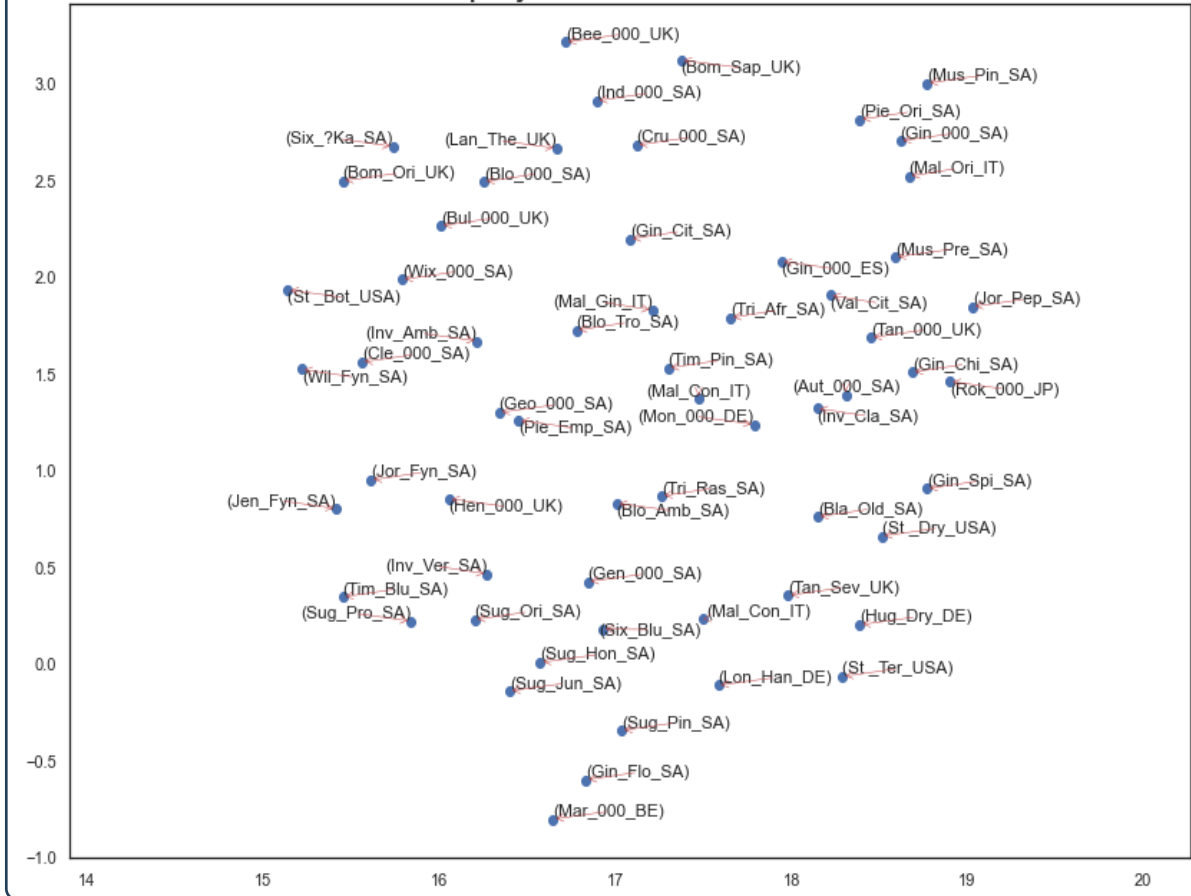
The application of this approach to small product sets:

- Evaluation of various module choices
- Optimization of
 - time
 - effort
 - cost

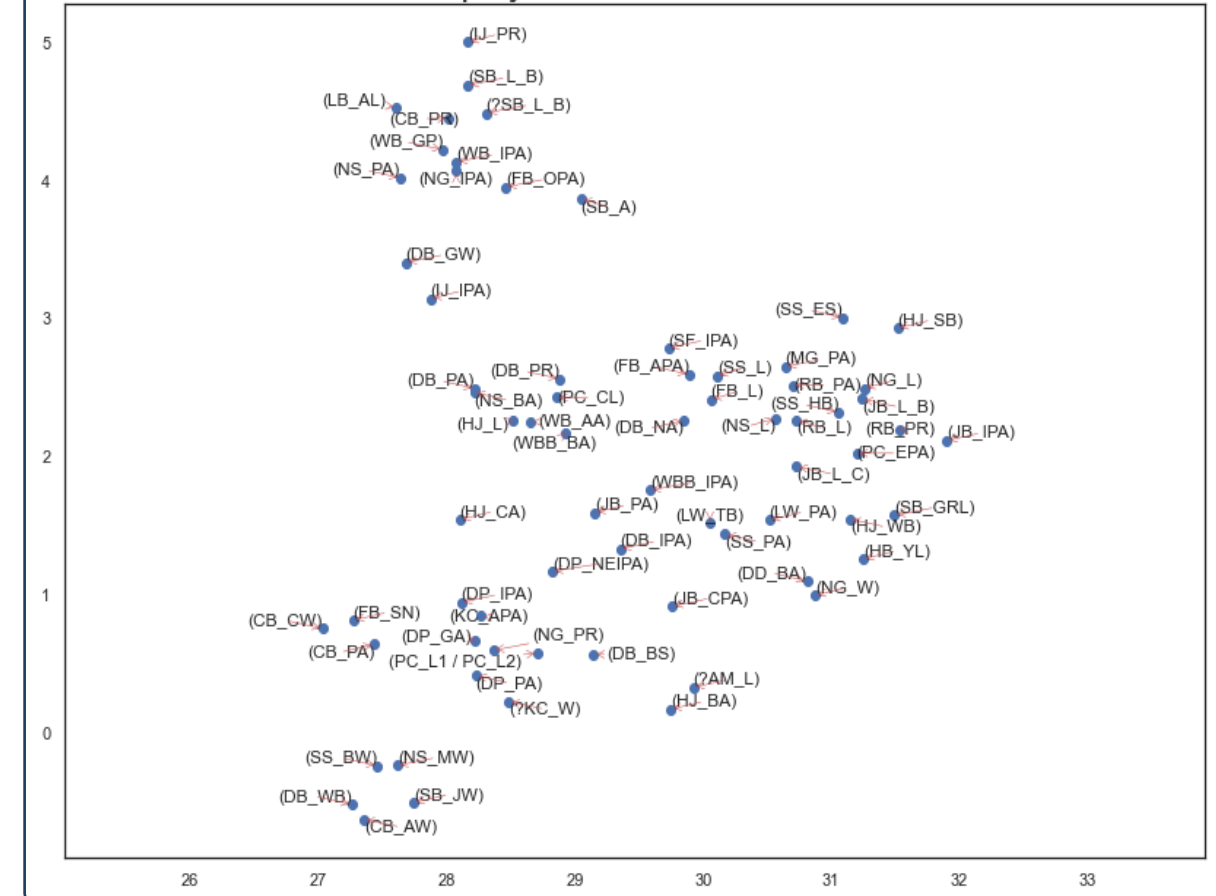


Application (gin and beer): Better visualization?

UMAP projection of the Gin dataset



UMAP projection of the Beer dataset



How can this be used?

- Once established, easy and cheap access to information
- Benefits for product placement and marketing = increased revenue
- Adds value to the development of future products where the desired properties can be assessed
- = better wine, beer, and gin (Chocolate, soft drinks, etc.)



What is the next product for investigation?

The optimized tool can be deployed in much larger projects like investigating the sensory space of Chenin Blanc wines, specifically looking at old vine and new vine comparisons (Requires rewriting of module for description sources) [SQL, Scrapy, Beautiful Soup, Power Automate]

The older the vine, the better the wine?



Thank you very much for your attention.
Questions are most welcome.



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