

# How spectrally representative are datasets used to build MIR-based predictive models ? A data-driven study.



C. Nickmilder<sup>1</sup>, J. Leblois<sup>2</sup>, O. Christophe<sup>3</sup>, C. Grelet<sup>3</sup>, Holicow Consortium, H. Soyeurt<sup>1</sup>

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ICAR meeting, bled, May 2024

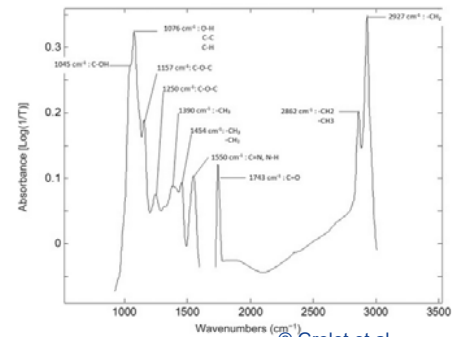
<sup>1</sup>Gembloux Agro-Bio Tech, University of Liège, Belgium

<sup>2</sup>Eleveo, Walloon Breeding Association, Ciney, Belgium

<sup>3</sup>Walloon Research Centre (CRA-W), Gembloux, Belgium



# Milk Recording



© Grelet et al., 2015

MIR-based equation



©Avelino Calvar Martinez



Consumption index, nitrogen efficiency ...

Methane, P, urea ...

Nutritional quality

Fat, protein, lactose, fatty acids, Ca, lactoferrin,...

Sustainability

Technological properties

Cheese yield, yoghurt yield, butter yield, spreadability ...



Environmental fingerprint

Animal Health

Na, lactoferrin, Energy balance, body weight, dry matter intake, acetone, BHB, citrate

Can we applied all MIR equations on those data as they were built on different calibration set ?

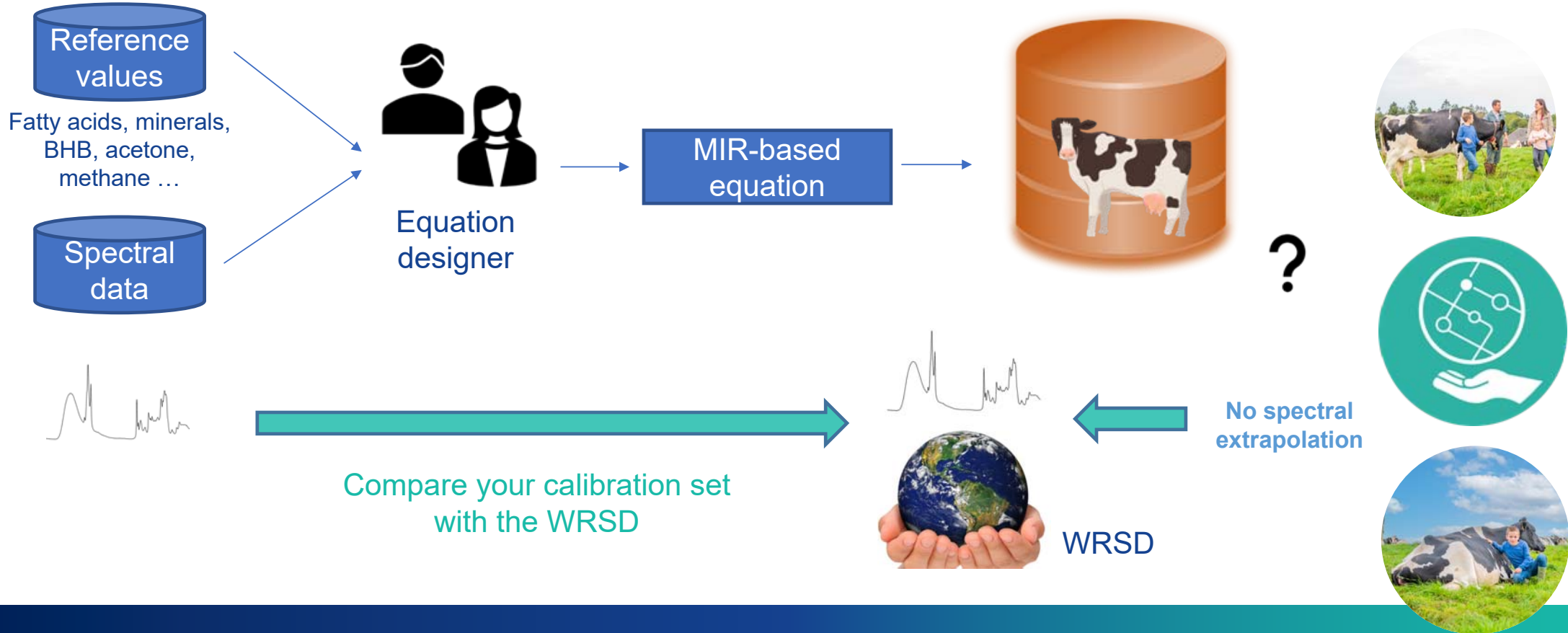


**+/- 63,000,000 spectra**  
**6 countries**  
**22 different breeds**

Partner	Country	N° spectra	N° herds	N° cows	Nbreed
Elevéo	BEL	5,813,993	1,508	317,674	11
Qualitas	CHE	3,672,804	1,890	214,404	8
LKV BW	DEU	2,137,394	503	99,814	15
LKV NRW	DEU	2,142,664	2,388	260,941	17
LKV SH	DEU	8,882,066	2,142	432,598	16
Eliance	FRA	38,353,951	20,824	2,555,698	22
ICBF	IRL	27,610	153	10,663	12
CONVIS	LUX	1,248,653	556	34,629	8



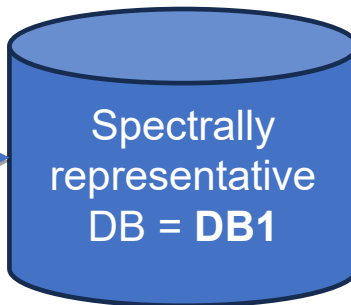
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# Spectral reduction



N = 52,303,184



WRSD reduction  
(Soyeurt et al., 2024)



Initially proposed in  
ICAR meeting, 25th May 2023

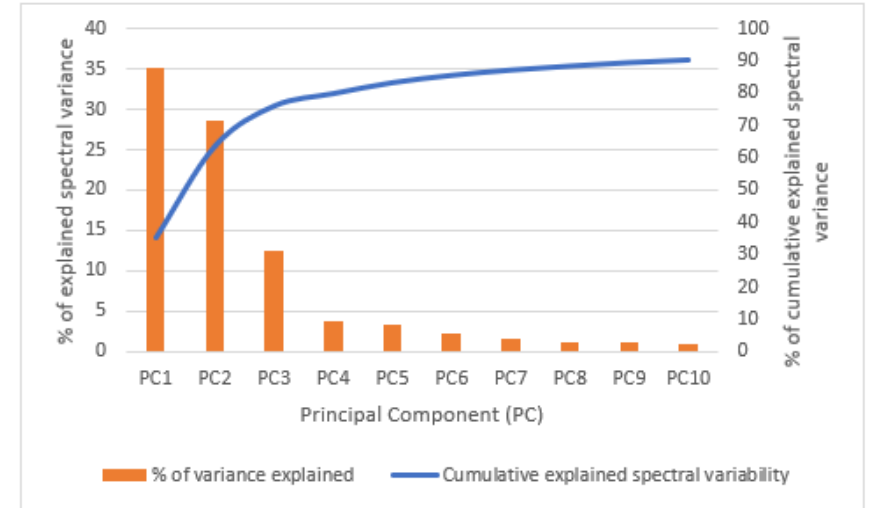
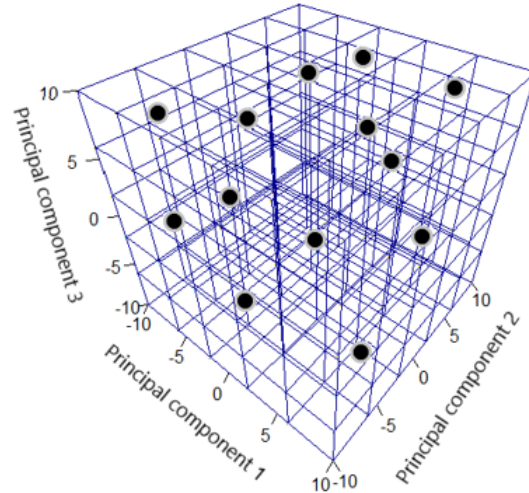
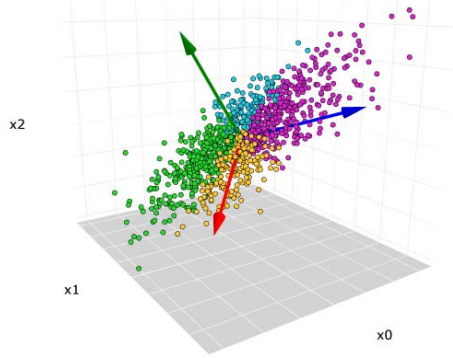
Update for  
ICAR meeting, 20th May 2024

# World representative spectral database (WRSD)

H. Soyeurt, C. Nickmilder, S. Franceschini, M. Whittaker, F. Dehareng, M. Bahadi, J. Leblois, L. Dale, K. Sanders, C. Grelet

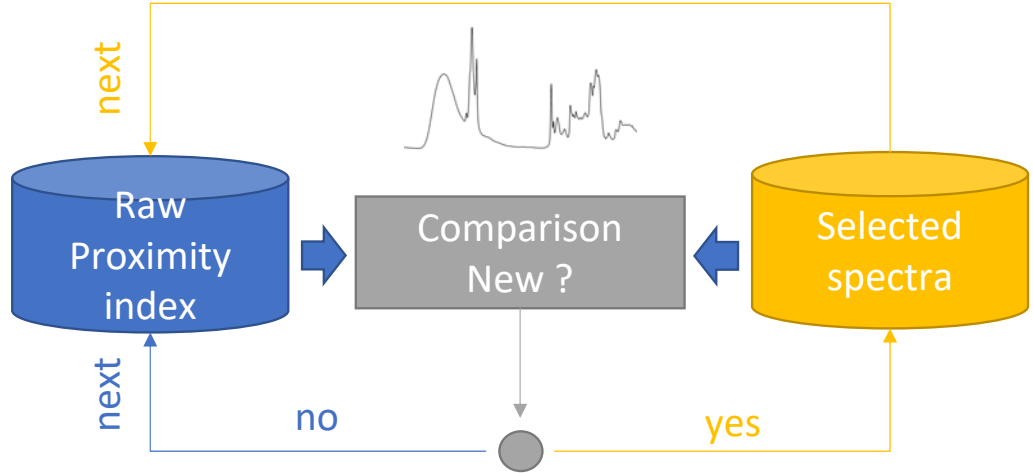
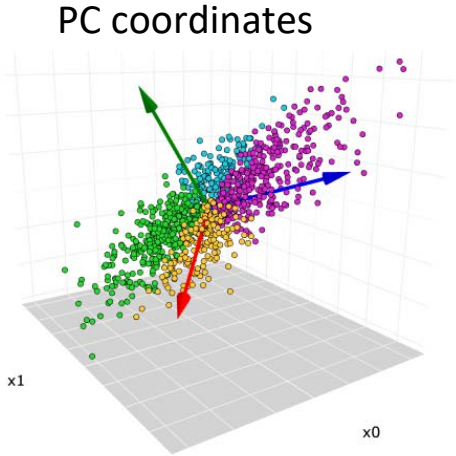
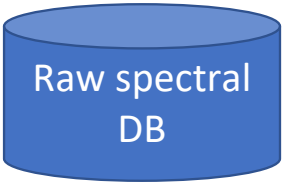
# Use the scores on the PC

- Scores on the Principal Components
- Create a grid based on the scores on PC
- Only one iteration



More PCs → More explained spectral variability

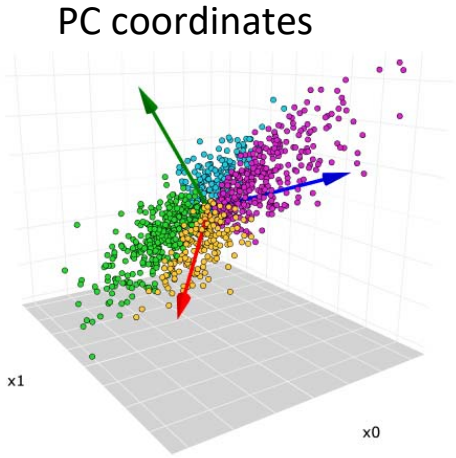
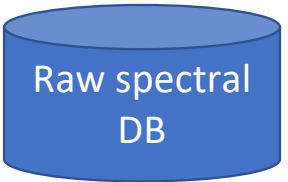




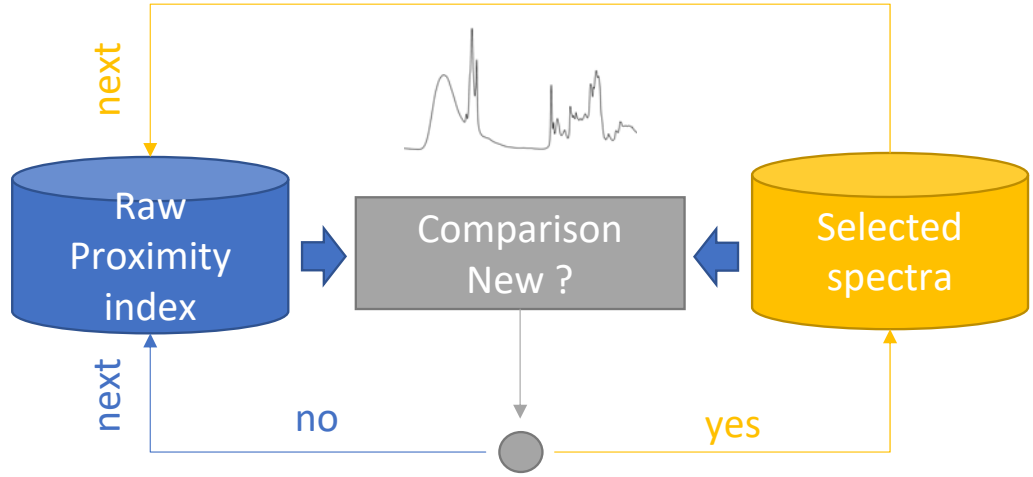
The code was written in Python in order to be run easily in each data center.

Proximity index :

•`Paste(round(PC1;0),round(PC2;0),round(PC3;0))`



Proximity index based on 3 PCs



Raw Spectral DB :

•Foss Database (DB)

- Eleveo Spectral DB : ± 8,000,000 records
- Lactanet Spectral DB : ± 10,000,000 records

•Bentley Database:

- LIC Spectral DB : ± 2,000,000 records
- LKV-BW Spectral DB : ± 10,000,000 records

Final Selection :

•Foss Database

- Eleveo Spectral DB : **167,015** records
- Lactanet Spectral DB : **172,469** records

•Bentley Database:

- LIC Spectral DB : **81,080** records
- LKV-BW Spectral DB : **91,494** records

An article was written about the methodology and submitted to the Journal of Dairy Science

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N = 52,303,184

Spectral reduction



Spectra subset DB1

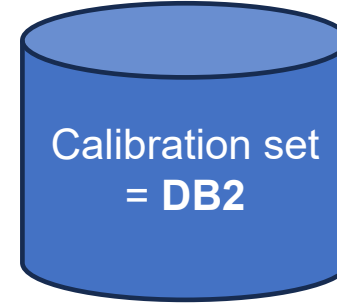


N = 322,216

Difference between 2 volumes



%coverage ?



N = 2,000  
Fatty Acids FT-MIR spectra

To conserve the information about the spectral distribution in the studied cow population, the frequency of the selected spectra is calculated



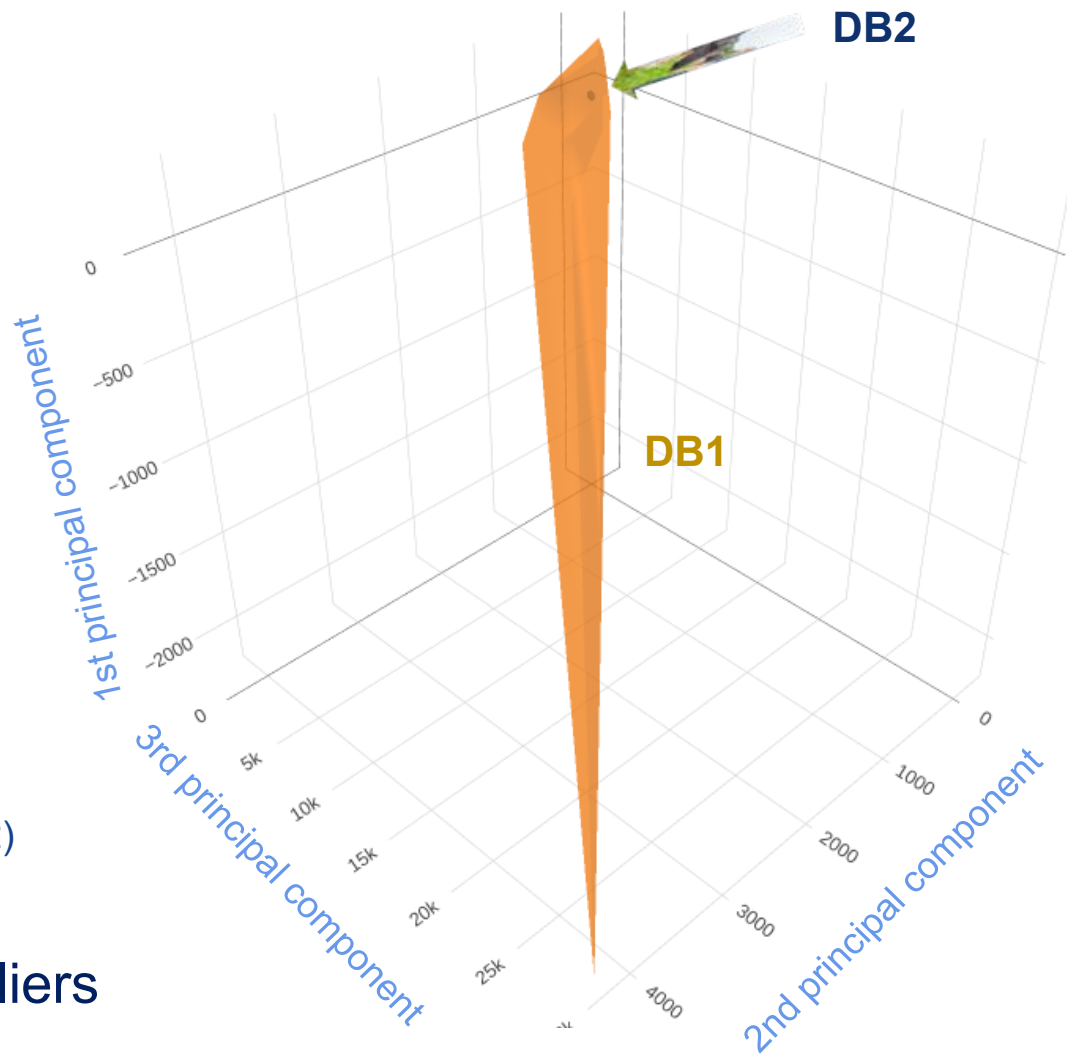
# Visualisation of the full Holicow subset (DB1)

Volume drawn by the subset

Volume drawn by the calibration set (DB2)



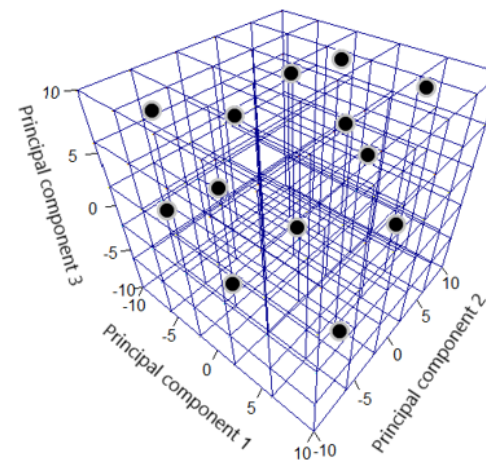
Presence of outliers



# Data Cleaning

- Standardized Mahalannobis distance < 5

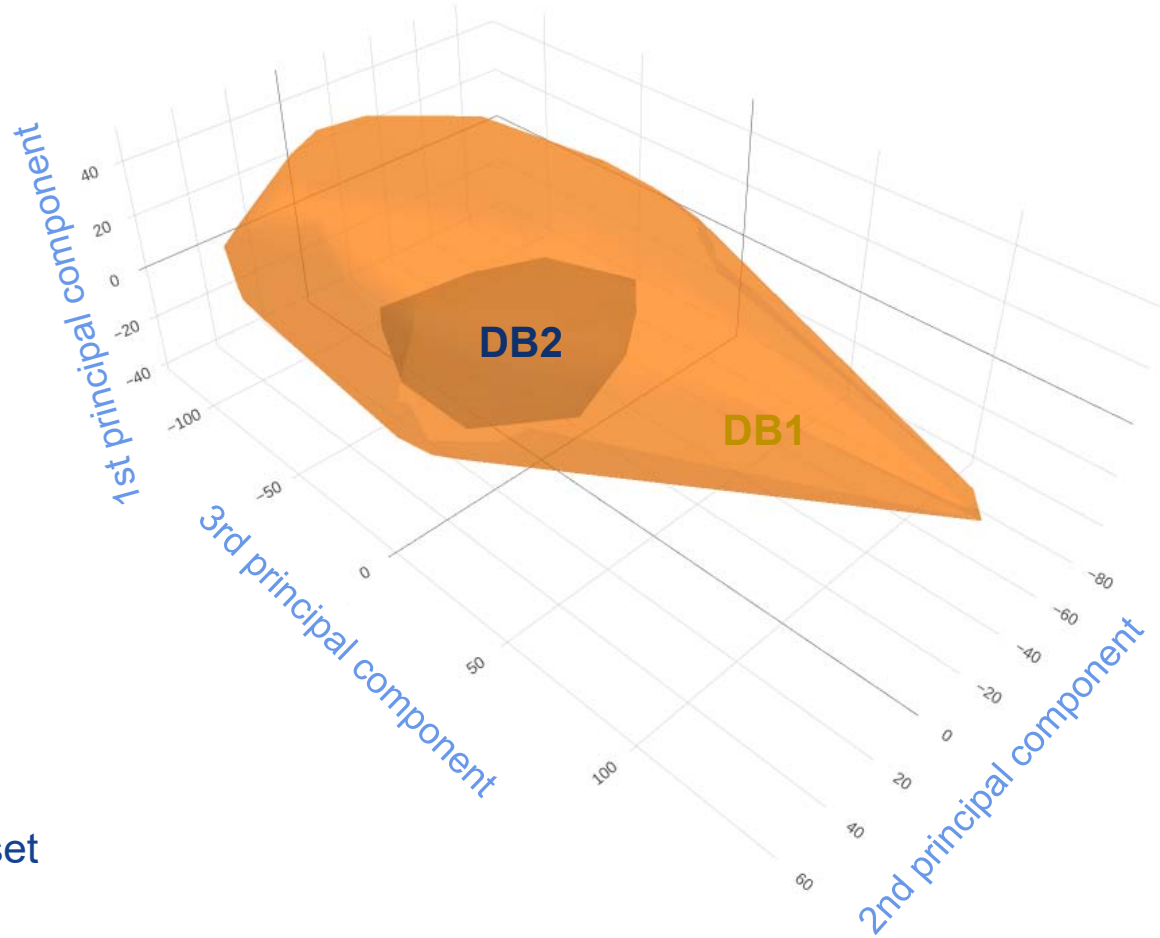
Cleaning	Holicow WRSD (DB1)	%kept
Raw	322,216	100%
Cleaned	275,593	85.53%



## Cleaned Holicow subset

Volume drawn by the subset

Volume drawn by the calibration set



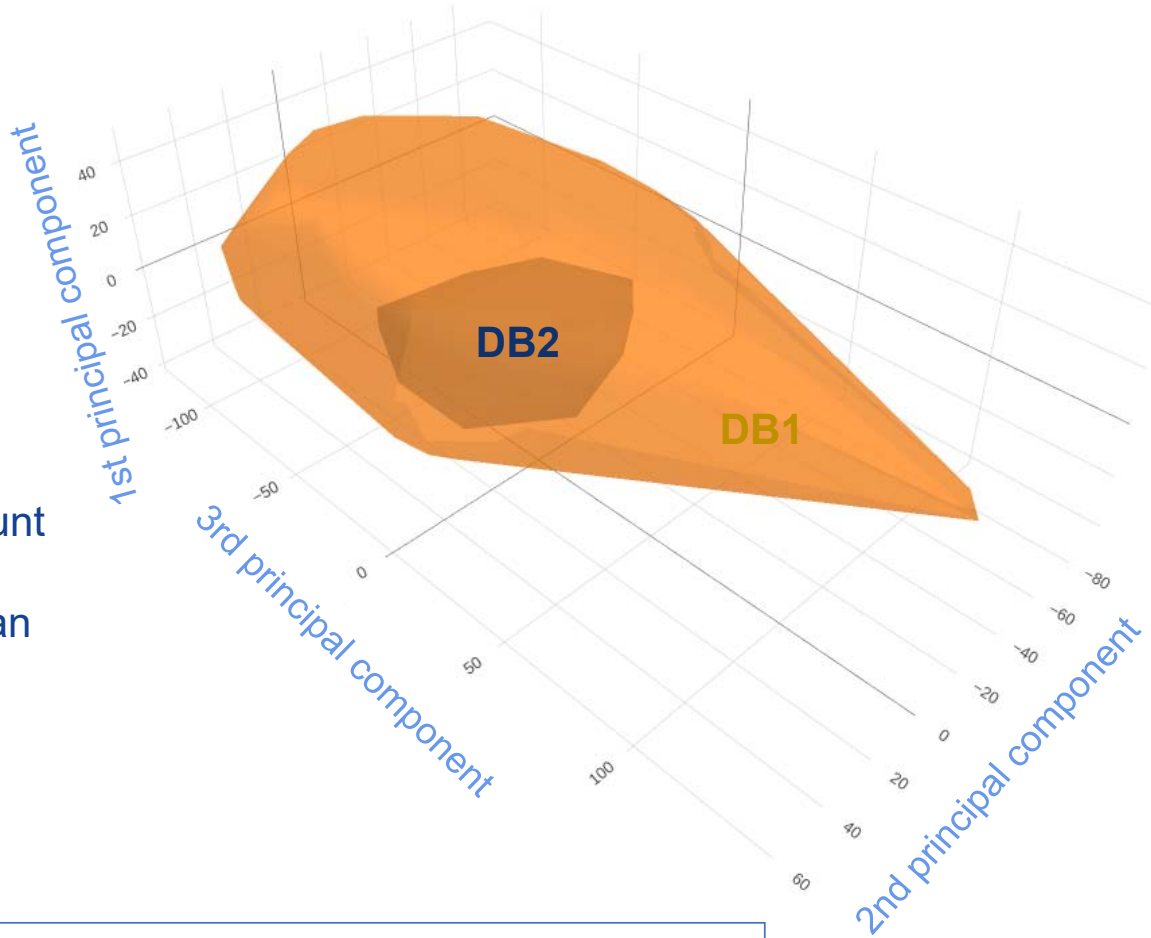
Without taken into account the density

DB2 Volume = 0.12 \* DB1 Volume

DB2 covers 99.64% of the spectral variability of DB1 if we take into account the density in the population

→ Some spectra are more present than others

→ More important to be in the area in which the frequency of spectra is the higher.



Fatty acids equations can be applied





## Conclusions

- WRSD method selects, as expected, outliers → **data cleaning**
- Need to consider the **frequency** of each spectra to have a correct conclusion.
- Even if the calibration set was limited, it has a variability that allow to cover the variability in the Holicow Database → **FA equations can be applied.**





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