

Zootechnical parameters added to the milk MIR spectra as predictive value to estimate CH₄ emissions

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Global Warming of 1.5 °C

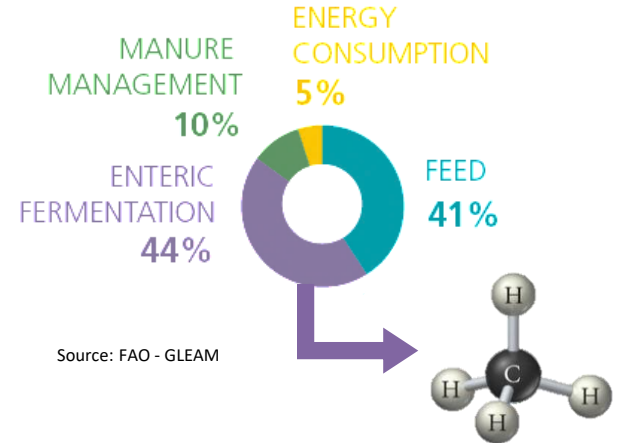
An IPCC special report on the impacts of global warming of 1.5 °C above pre-Industrial levels and related global greenhouse gas emission pathways, in the context of strengthening the global response to the threat of climate change, sustainable development, and efforts to eradicate poverty.

Context



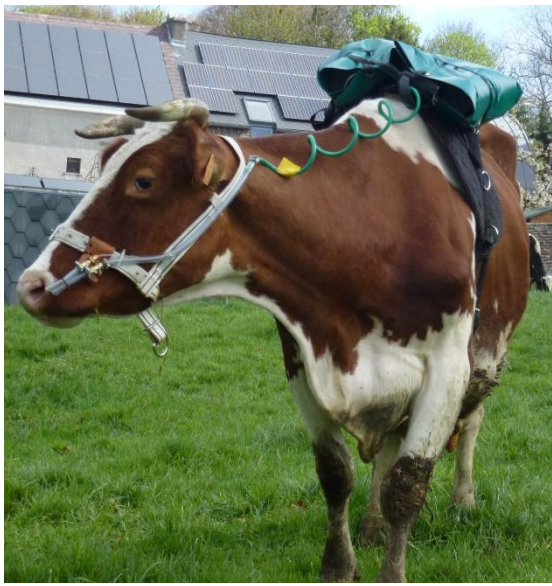
→ Efforts from each sector to reduce GHG emissions

→ Importance of enteric fermentation in agriculture →



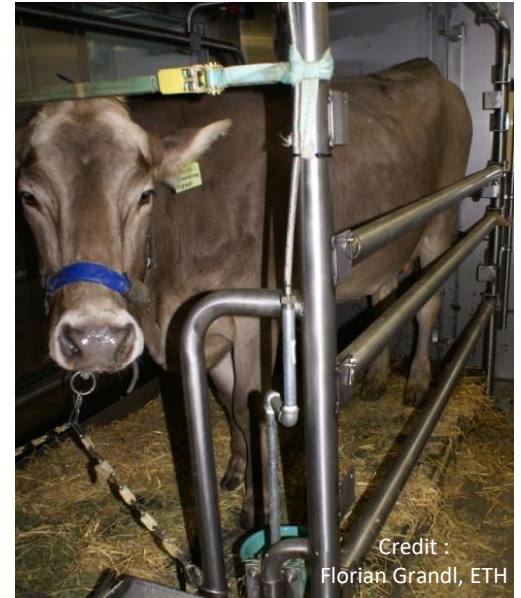
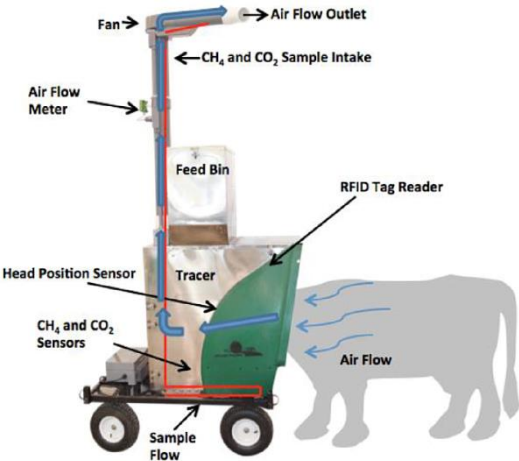
→ CH₄ measurements to establish the link with - diet

- genetics → Large scale studies

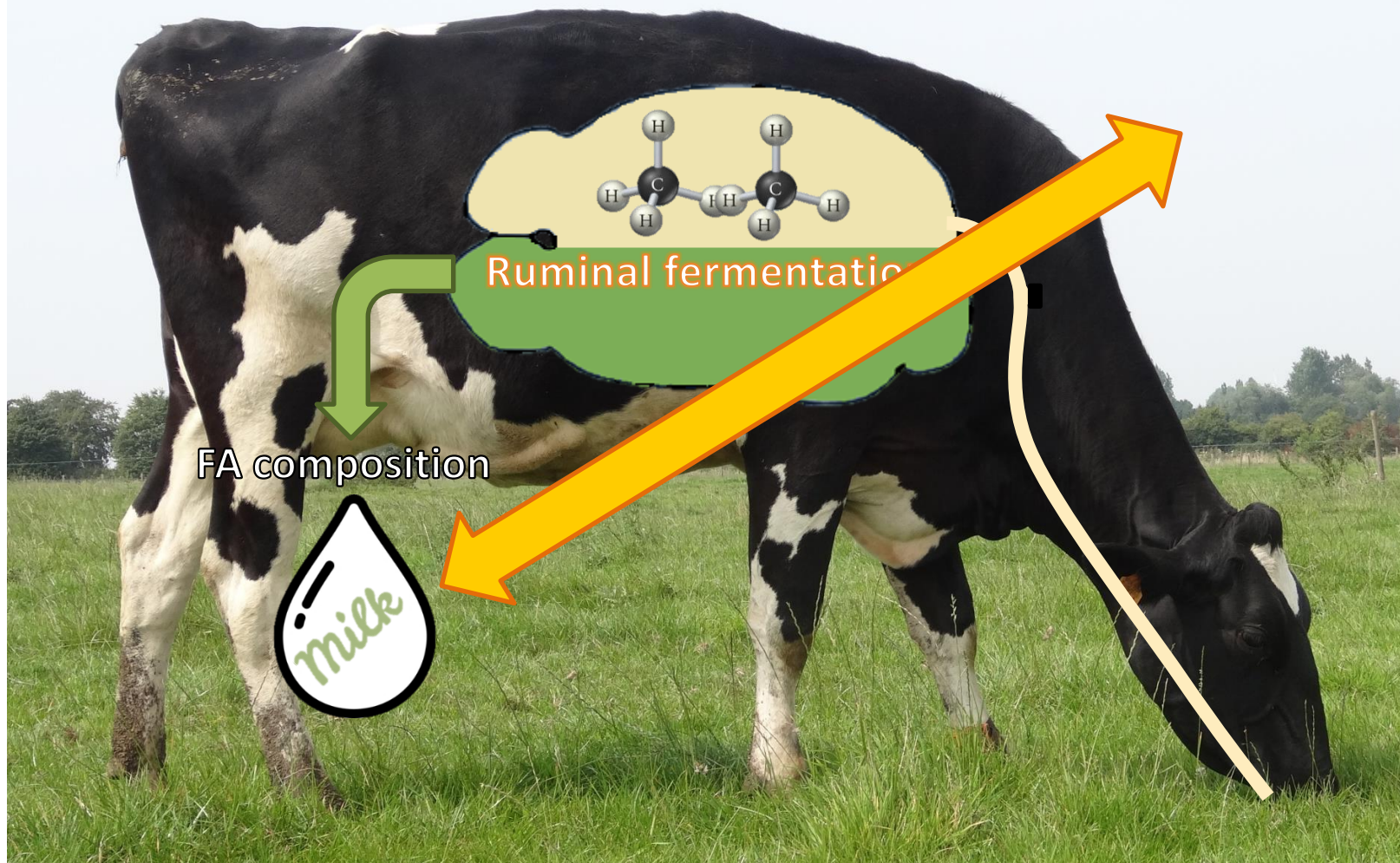


Credit :
Jan Lassen, Aarhus University

→ Need of a proxy
to consider individual variability



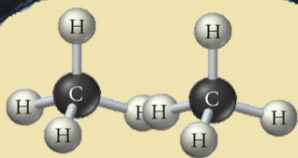
Credit :
Florian Grandl, ETH



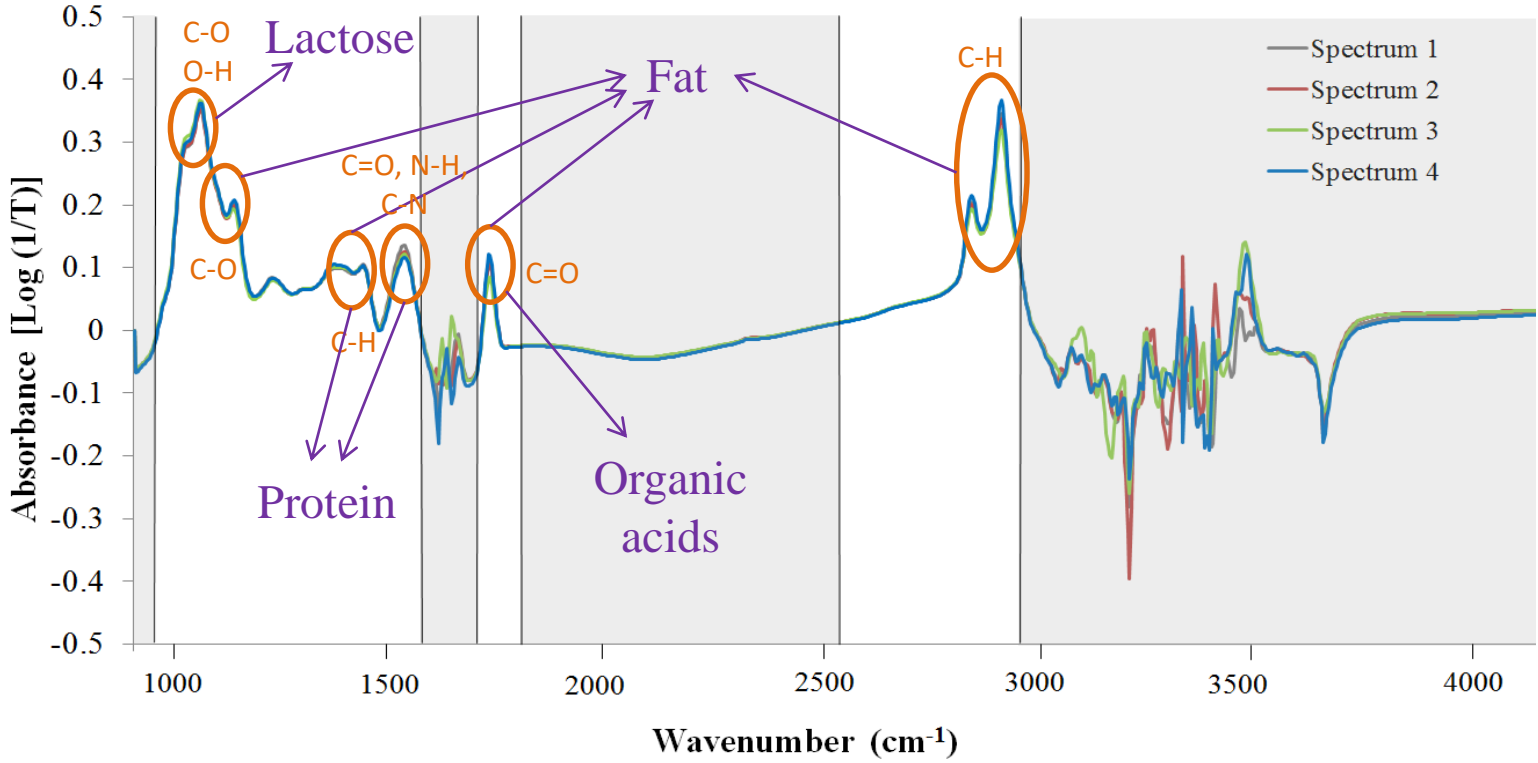
FA composition



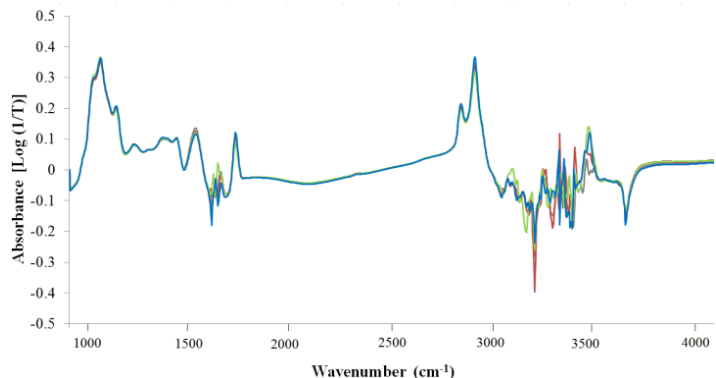
Ruminal fermentation



Milk FT-MIR spectra as a proxy for enteric CH₄



Development of equations to estimate CH₄ from milk FT-MIR spectra



However...

Huge difference between

Developing a model in a research context

Optimized performances (highest R², RMSE)
Limited variability
Complex models



Using a model to generate predictions at a large scale








Optimized robustness
Great variability
Simple models (#IV, WN, methods)



→ Consideration of a maximum of variability

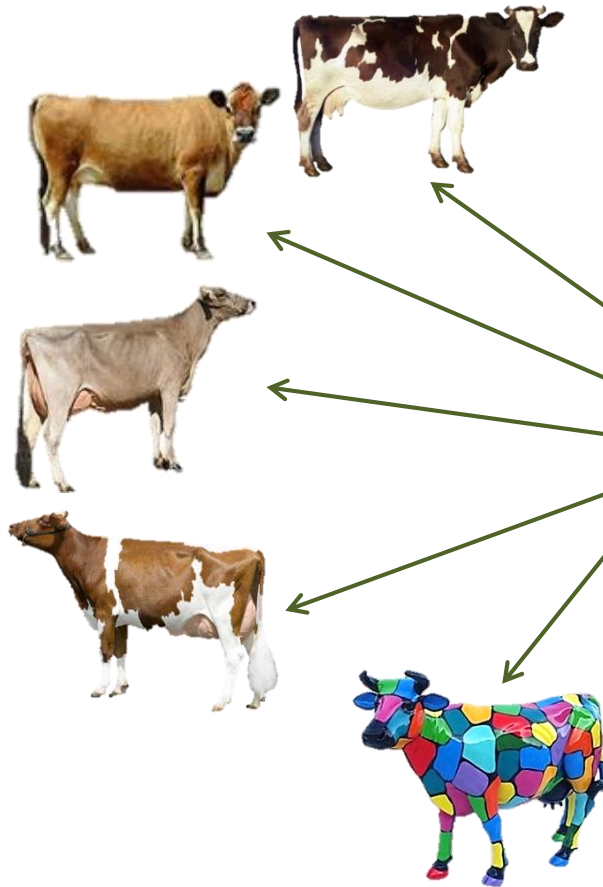
Description of the reference data



		CH ₄ (g/day)	N data	N cows
SF ₆		510 ± 105	252	42
		347 ± 89	261	110
	TOTAL	427 ± 127	513	152
Chambers		405 ± 60	207	50
		451 ± 75	135	57
		367 ± 64	129	19
		366 ± 61	81	9
		365 ± 44	24	12
	TOTAL	400 ± 72	576	147
TOTAL		413 ± 102	1,089	299

Description of the reference data

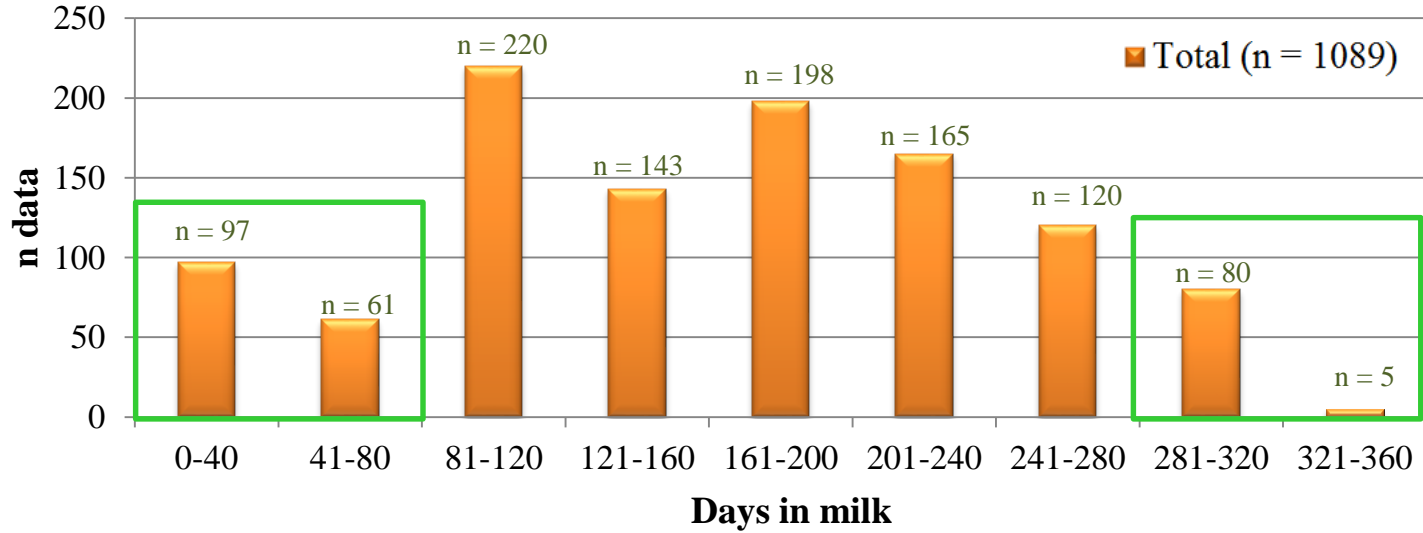
→ Breeds



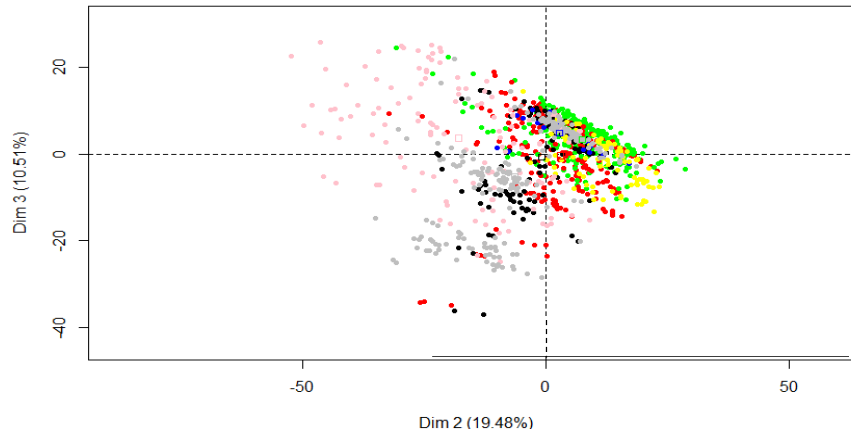
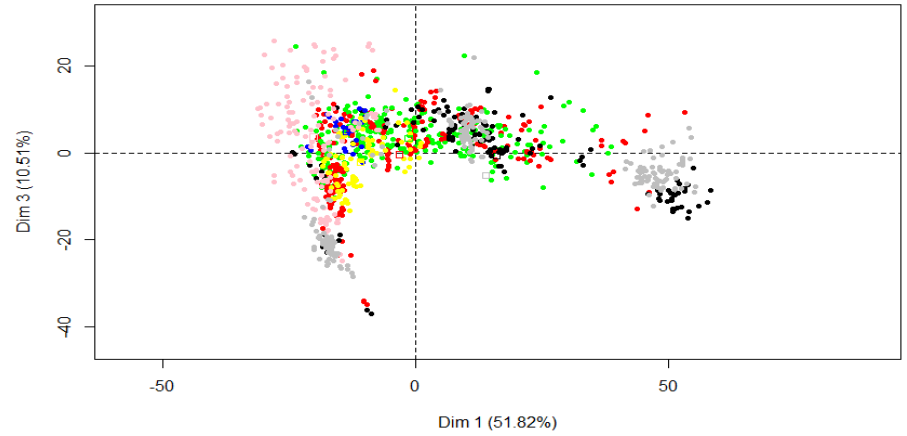
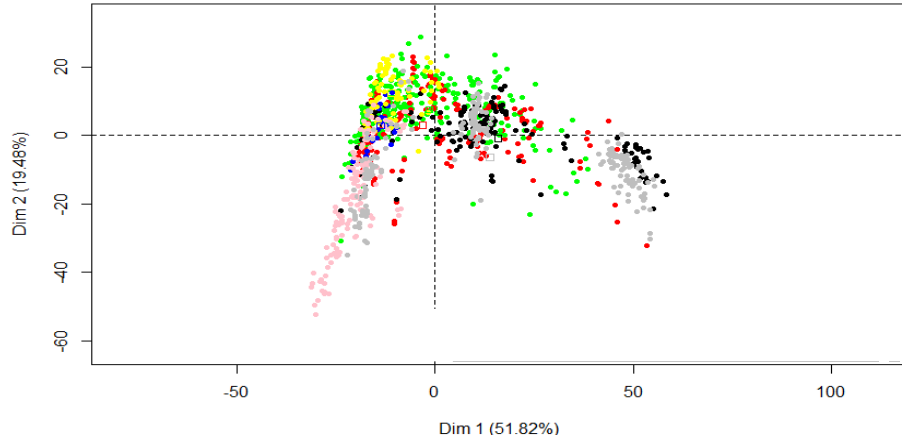
Breed	n data	% of data	n cows	% of cows	CH ₄ (g/d) mean ± SD
HO	891	82	222	74	415 ± 107
JER	67	6	10	3	342 ± 42
BS	78	7	39	13	458 ± 69
RH	21	2	8	3	427 ± 74
X	32	3	20	7	391 ± 67

Description of the reference data

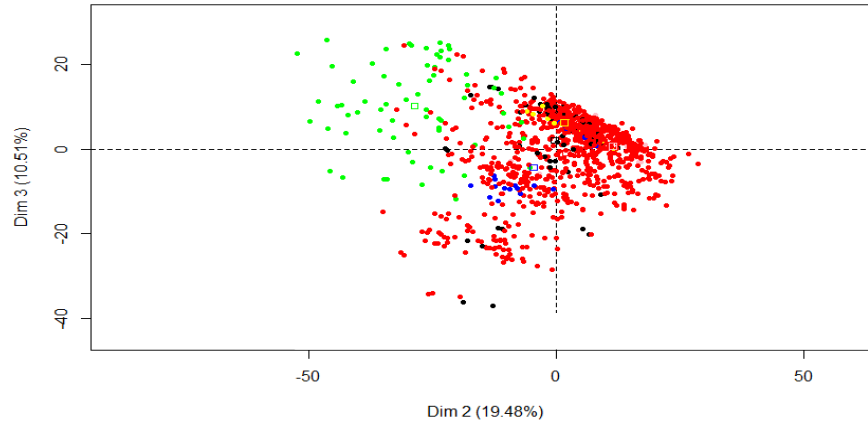
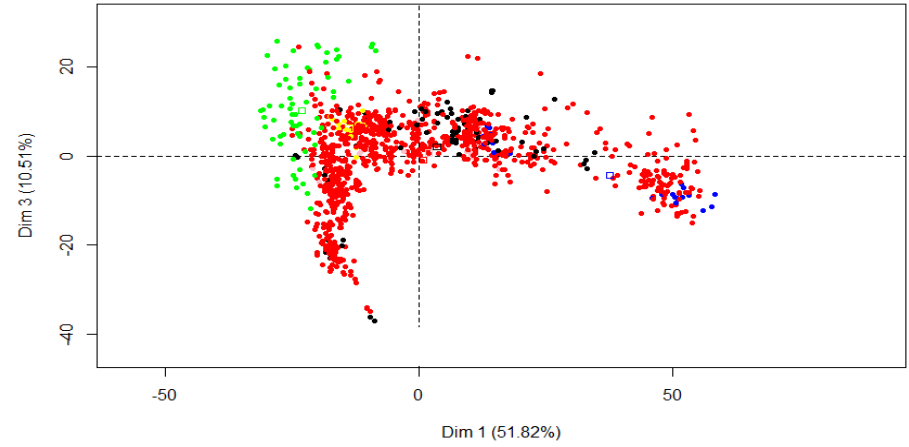
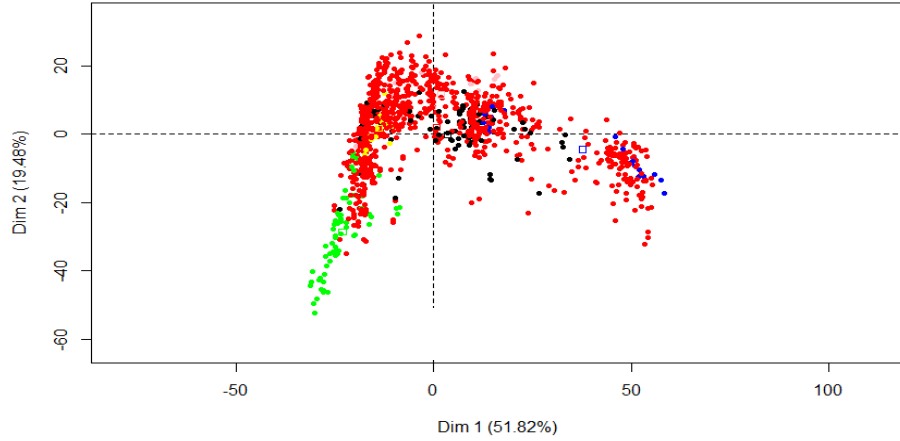
→ Lactation stage



Description of the reference data → FT-MIR spectra



Description of the reference data → FT-MIR spectra



Methods

Milk MIR spectra



Methods



Milk MIR spectra

Standardization



J. Dairy Sci. 100:1–12
<https://doi.org/10.3168/jds.2017-12720>
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Standardization of milk mid-infrared spectrometers for the transfer and use of multiple models

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Integration of the “Days In Milk” (modified Legendre Polynomials)



J. Dairy Sci. 98:5740–5747
<http://dx.doi.org/10.3168/jds.2014-8436>
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Hot topic: Innovative lactation-stage-dependent prediction of methane emissions from milk mid-infrared spectra

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Methods

Methane measurements

→ Raw data

Methods

Calibration & validations

→ WinISI 4.6 (Foss)

→ PLS

→ 5 groups cross-validation

Randomly 4/5 of data to calibrate, 1/5 to validate (all groups combinations)

→ Cow and Country Dependent External Validation (CCDEV)

20% of cows randomly removed per country

Calibration with the other 80% of cows, validation on the 20%

Loop of 500 repetitions

Equation developped

	CH₄ Ref. method	n data	n cows	Origin	Pred. variables	R²c	SEC (g/d)	R²cv	SECV (g/d)
Prediction Model	SF ₆ & RC	1089	299	BE, IE, CH, UK, FR, DK, DE	MS	0.68	58	0.64	61

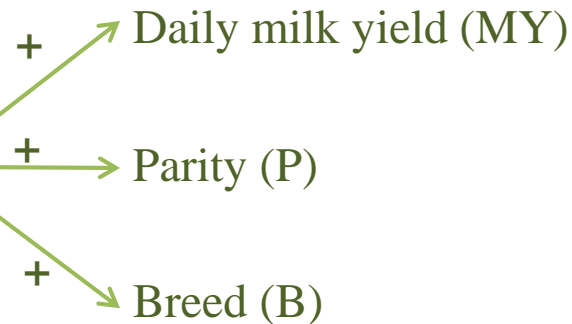
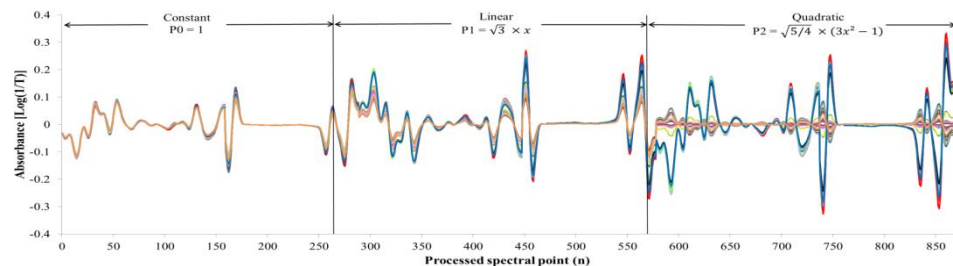
	R²_{CCDEV}	RMSEP_{CCDEV}
Prediction model	0.55 ± 0.07	70 ± 4.5

Inclusion of zootechnical information as predictive value



Information easily available

Modified milk FT-MIR spectra (MS)



Individually
&
combined

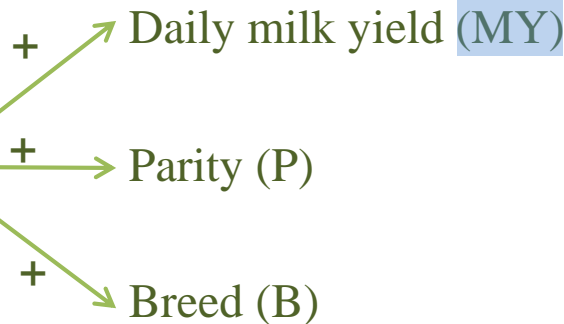
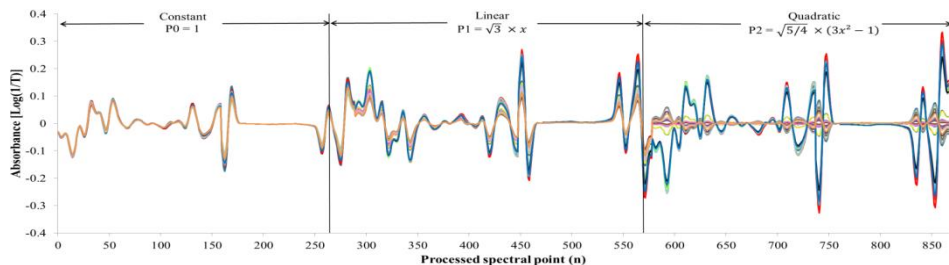
ID	Spectra				
	Abs 1	Abs 2	...	Abs 866	Abs 867
Cow 1	-0.07288	-0.06950	...	0.03138	0.03286
Cow 2	-0.09036	-0.08670	...	0.02072	0.02133
Cow 3	-0.08454	-0.08089	...	0.01950	0.01944
Cow 4	-0.08449	-0.08105	...	0.02084	0.01981
Cow 5	-0.08657	-0.08325	...	0.01589	0.01754

Inclusion of zootechnical information as predictive value



Information easily available

Modified milk FT-MIR spectra (MS)



Individually & combined

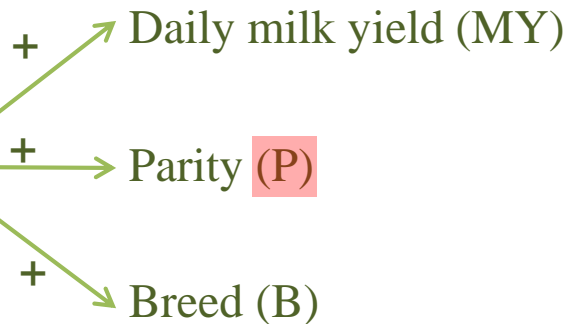
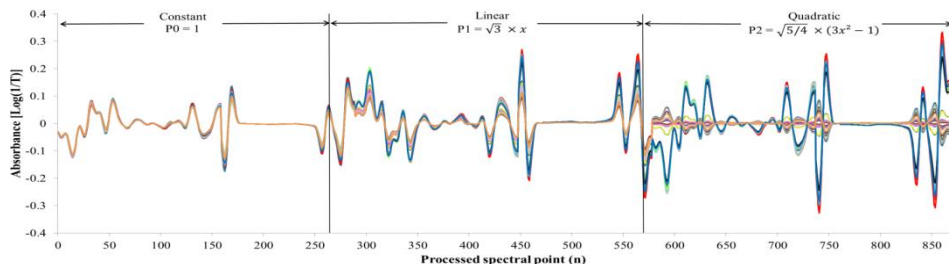
ID	MY Abs 1	Spectra		Spectra	
		Abs 12	Abs 2	Abs 866	Abs 867
Cow 1	-0.07288	-0.07288	-0.06950	0.03138	0.03286
Cow 2	-0.09036	-0.09036	-0.08670	0.02072	0.02133
Cow 3	-0.08454	-0.08454	-0.08089	0.01950	0.01944
Cow 4	-0.08449	-0.08449	-0.08105	0.02084	0.01981
Cow 5	-0.08657	-0.08657	-0.08325	0.01589	0.01754

Inclusion of zootechnical information as predictive value



Information easily available

Modified milk FT-MIR spectra (MS)



Individually
&
combined

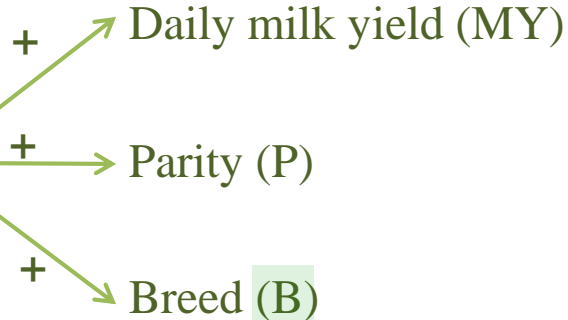
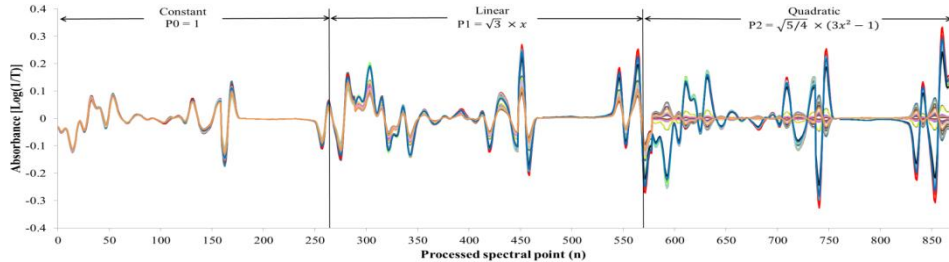
ID	MPY	Spectra				
		Abs 1	Abs 2	...	Abs 866	Abs 867
Cow 1	24	-0.07288	-0.06950	...	0.03138	0.03286
Cow 2	19	-0.09036	-0.08670	...	0.02072	0.02133
Cow 3	23	-0.08454	-0.08089	...	0.01950	0.01944
Cow 4	26	-0.08449	-0.08105	...	0.02084	0.01981
Cow 5	20	-0.08657	-0.08325	...	0.01589	0.01754

Inclusion of zootechnical information as predictive value



Information easily available

Modified milk FT-MIR spectra (MS)



Individually & combined

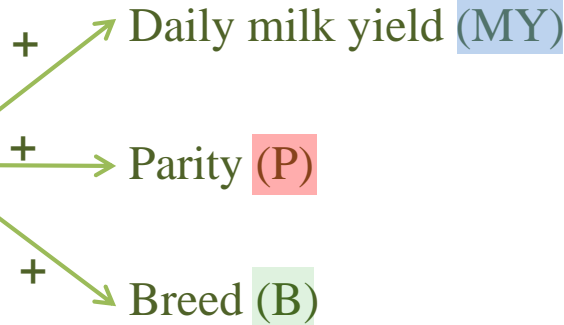
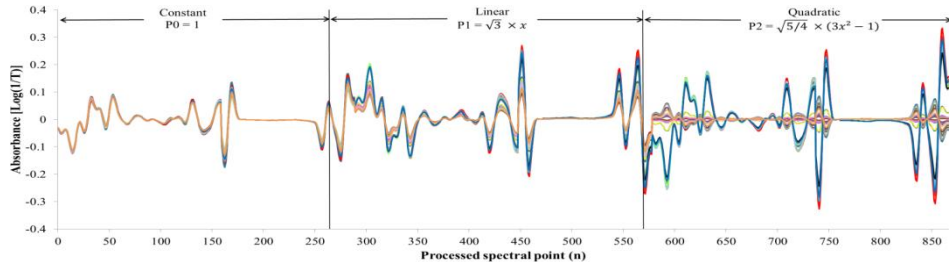
ID	Spectra					Spectra				
	Breed 1 P	Breed 2 Abs 1	Breed 3 Abs 2	Breed 4 ...	Other Breed	Abs 867	Abs 2	...	Abs 866	Abs 867
Cow 1	1	-0.07288	-0.06950	..0	0.03188	0.03288	-0.06950	...	0.03138	0.03286
Cow 2	2	-0.09036	-0.08670	..0	0.02072	0.02136	-0.08670	...	0.02072	0.02133
Cow 3	3	-0.08454	-0.08089	..0	0.01950	0.01944	-0.08089	...	0.01950	0.01944
Cow 4	3	-0.08449	-0.08105	..0	0.02084	0.01981	-0.08105	...	0.02084	0.01981
Cow 5	1	-0.08657	-0.08325	..0	0.01589	0.01754	-0.08325	...	0.01589	0.01754

Inclusion of zootechnical information as predictive value



Information easily available

Modified milk FT-MIR spectra (MS)



Individually & combined

IDD	MY	Breed 1	IP	Breed 2						Other Breed		Spectra		Spectra			
				Breed 2	Breed 2	Breed 2	Breed 2	Breed 2	Breed 2	Breed 3	Breed 4	Abs 1	Abs 2	Abs 866	Abs 867	Abs 867	
C6v1	1	24	1	1	0	1	0	0	0	0	0	0	-0.07288	-0.06950	0.03438	0.03286	0.03286
C6v2	2	19	0	2	0	0	1	0	0	1	0	0	-0.09036	-0.08670	0.02072	0.02133	0.02133
C6v3	3	21	0	3	1	0	0	1	0	0	0	0	-0.08454	-0.08089	0.01950	0.01944	0.01944
C6v4	4	26	1	3	0	1	0	0	0	0	0	0	-0.08449	-0.08105	0.02084	0.02181	0.01981
C6v5	5	20	1	1	0	1	0	0	0	0	0	0	-0.08657	-0.08325	0.01589	0.01754	0.01754

Equations developped

	CH ₄ Ref. method	n data	n cows	Origin	Pred. variables	R ² c	SEC (g/d)	R ² cv	SECV (g/d)
Prediction Model	SF ₆ & RC	1,089	299	BE, IE, CH, UK, FR, DK, DE	MS	0.68	58	0.64	61
Inclusion of zootechnical information	SF ₆ & RC	1,089	299	BE, IE, CH, UK, FR, DK, DE	MS+MY	0.69	57	0.65	60
					MS+P	0.70	56	0.65	60
					MS+B	0.70	56	0.66	60
					MS+MY+P	0.70	56	0.66	60
					MS+MY+B	0.72	54	0.68	58
					MS+P+B	0.71	55	0.67	59
					MS+MY+P+B	0.73	53	0.68	57

Equations developped

	CH ₄ Ref. method	n data	n cows	Origin	Pred. variables	R ² c	SEC (g/d)	R ² cv	SECV (g/d)
Prediction Model	SF ₆ & RC	1,089	299	BE, IE, CH, UK, FR, DK, DE	MS	0.68	58	0.64	61
Inclusion of zootechnical information	SF ₆ & RC	1,089	299	BE, IE, CH, UK, FR, DK, DE	MS+MY	0.69	57	0.65	60
					MS+P	0.70	56	0.65	60
					MS+B	0.70	56	0.66	60
					MS+MY+P	0.70	56	0.66	60
					MS+MY+B	0.72	54	0.68	58
					MS+P+B	0.71	55	0.67	59
					MS+MY+P+B	0.73	53	0.68	57

Equations developped

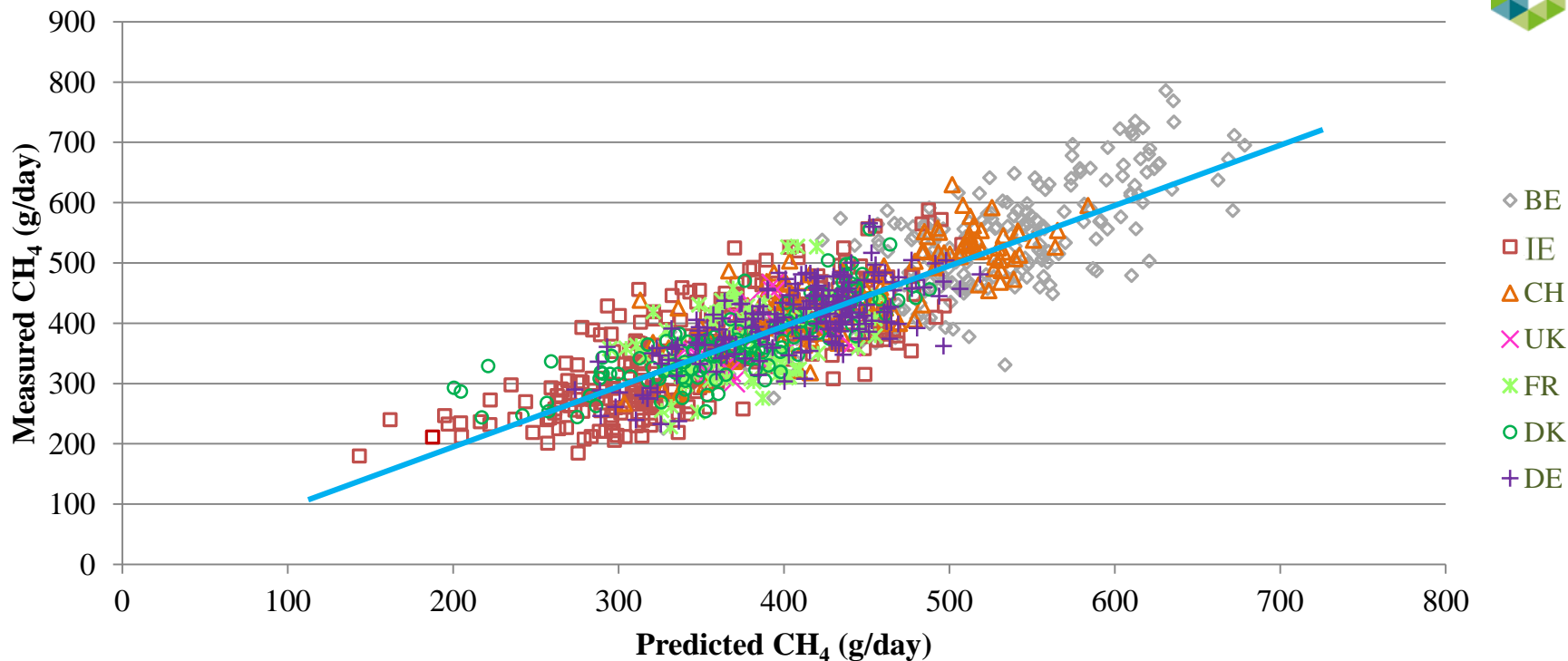
	CH ₄ Ref. method	n data	n cows	Origin	Pred. variables	R ² c	SEC (g/d)	R ² cv	SECV (g/d)
Prediction Model	SF ₆ & RC	1,089	299	BE, IE, CH, UK, FR, DK, DE	MS	0.68	58	0.64	61
Inclusion of zootechnical information	SF ₆ & RC	1,089	299	BE, IE, CH, UK, FR, DK, DE	MS+MY	0.69	57	0.65	60
					MS+P	0.70	56	0.65	60
					MS+B	0.70	56	0.66	60
					MS+MY+P	0.70	56	0.66	60
					MS+MY+B	0.72	54	0.68	58
					MS+P+B	0.71	55	0.67	59
					MS+MY+P+B	0.73	53	0.68	57

Inclusion of zootechnical information as predictive value



Model	Effect on the error of prediction		
	Parity	Country	Breed
MS	***	***	***
MS + MY	***	***	***
MS + P	NS	**	***
MS + B	***	NS	***
MS + MY + P	NS	***	***
MS + MY + B	***	NS	**
MS + P + B	NS	NS	*
MS + MY + P + B	NS	NS	*

Equations developed



	CH ₄ Ref. method	n data	n cows	Origin	Pred. variables	R ² c	SEC (g/d)	R ² cv	SECV (g/d)
Inclusion of zootechnical information	SF ₆ & RC	1089	299	BE, IE, CH, UK, FR, DK, DE	MS+MY+P+B	0.73	53	0.68	57

Practical applications

Walloon milk recording Jan. 2016 → Dec. 2017

5 – 365 DIM → n = 929,966

Model	n GH > 5	% GH > 5
MS	52,344	5.6
MS + MY	50,994	5.5
MS + P	50,034	5.4
MS + B	183,454	19.7
MS + MY + P	49,529	5.3
MS + MY + B	143,622	15.4
MS + P + B	150,025	16.1
MS+ MY + P + B	120,637	13.0

Practical applications

Walloon milk recording Jan. 2016 → Dec. 2017

5 – 365 DIM → n = 929,966

Model	HO n = 618,743		JER n = 1,342		BS n = 4,559		RH n = 305		OTHER n = 305,017	
	n GH > 5	% GH > 5	n GH > 5	% GH > 5	n GH > 5	% GH > 5	n GH > 5	% GH > 5	n GH > 5	% GH > 5
MS	33,974	5.5	185	13.8	245	5.4	22	7.2	17,918	5.9
MS + MY	32,743	5.3	184	13.7	246	5.4	22	7.2	17,799	5.8
MS + P	32,459	5.2	183	13.6	238	5.2	22	7.2	17,132	5.6
MS + B	19,202	3.1	232	17.3	215	4.7	104	34.1	163,701	53.7
MS + MY + P	31,711	5.1	178	13.3	232	5.1	20	6.6	17,388	5.7
MS + MY + B	18,211	2.9	230	17.1	198	4.3	86	28.2	124,897	40.9
MS + P + B	18,888	3.1	230	17.1	217	4.8	115	37.7	130,575	42.8
MS + MY + P + B	18,469	3.0	237	17.7	204	4.5	100	32.8	101,627	33.3

Practical applications

Walloon milk recording Jan. 2016 → Dec. 2017

5 – 365 DIM → $n = 929,966$

→ $\text{GH} < 5$ → 248,982 spectra removed (67,6% from “OTHER”)

→ Holstein cows in RW

→ $150 \text{ g/d} < \text{predicted CH}_4 < 950 \text{ g/d}$ (0.12% of the remaining HO records)

→ $n = 538,510$

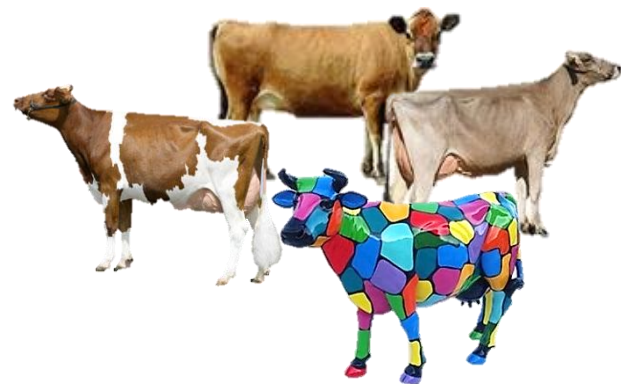
Perspectives for further improvements



- Majority of Holstein reference (82%) → Other breeds ?



- Increase the reference data
- 1 model/breed?



- Consideration of other relevant parameters as predictors

- Diet type, DMI, etc. → Practical applications?



Conclusions



- Milk MIR spectra is a great proxy for large scale studies related to CH₄
- Include MY, P or/and B
 - improve the R² and reduce the error of prediction
 - limit external impact on the error of prediction
- Importance to consider the conditions of application
- Need to investigate these models in practical conditions

Thank you for your attention