

Individual methane prediction from milk MIR spectra, across multiple breeds, lactation stages, partities and country-specific dairy farming systems



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Walloon Agricultural Research Centre

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#### **Context : Methane produced by ruminants**

Greenhouse gas + loss of gross energy intake (6 to 12%)

Sources of mitigation of CH<sub>4</sub> emissions :

- genetics

- diet
- management



Development of a technique that allows large scale studies

FT-IR used in Milk Recording Organisation ?



#### **Potential use of Mid infrared spectra of the milk ?**



## Position of the peaks → Qualitative analysis Intensity of the peaks → Quantitative analysis



#### **Potential use of Mid infrared spectra of the milk ?**





#### **Potential use of Mid infrared spectra of the milk ?**



#### **Innovative use of MIR spectra**





#### **Principle of prediction**





#### **First equation of prediction**



## Potential use of milk mid-infrared spectra to predict individual methane emission of dairy cows

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#### **First equation of prediction**



Figure 3 Infrared methane prediction on the basis of milk spectra of the day 1.5 for the different diets: corn silage ( $\bullet$ ), fresh pasture ( $\bigcirc$ ) and grass silage (+). PCA = principal component analysis.

| Equation  |  | SD  | R <sup>2</sup> c | R <sup>2</sup> cv | SEC  | SECV  |  |
|---|--|-----|------------------|-------------------|------|-------|--|
| CH <sub>4</sub> (g of CH <sub>4</sub> /Kg of milk)  |  | 7.3 | 0.85             | 0.75              | 4.25 | 5.61  |  |
| CH <sub>4</sub> (g/day)   |  | 128 | 0.77             | 0.70              | 87.8 | 100.6 |  |
| N = number of observations; SD = standard deviation; $R^2c$ = calibration coefficient of determination; $R^2cv$ = cross-validation coefficient of determination; SEC = calibration standard error; SECV = cross-validation standard error |  |     |                  |                   |      |       |  |





#### Merging of reference data sets

More data are needed to - include more variability

- improve performance of the equation



#### Merging of reference data sets

Use of different instruments







Standardisation step needed

EMR procedure (OptiMIR Project)





#### Merging of reference data sets







## Methane predictions depending on lactation stage



#### $\rightarrow$ Reversed curves

 $\rightarrow$  Need to improve our model



#### **Methane predictions** depending on lactation stage



3821 Victoria, Australia

CRA-W

Individual methane prediction from milk MIR spectra www.cra.wallonie.be

## Methane predictions depending on lactation stage

| Equation (g/day)        | Ν   | SD    | R <sup>2</sup> c | R <sup>2</sup> cv | SEC | SECV |
|-------------------------|-----|-------|------------------|-------------------|-----|------|
| CH <sub>4</sub>         | 446 | 132.6 | 0.78             | 0.74              | 63  | 68   |
| CH <sub>4</sub> and DIM | 446 | 127.5 | 0.75             | 0.67              | 63  | 72   |

N = number of observations; SD = standard deviation;  $R^2c$  = calibration coefficient of determination;  $R^2cv$  = cross-validation coefficient of determination; SEC = calibration standard error; SECV = cross-validation standard error

→ Statistical parameters are a slighty lower...

...BUT!



# Methane predictions depending on lactation stage

Application of CH<sub>4</sub> equations on Belgian spectral database 1<sup>st</sup> lactation Holstein cows



 $\rightarrow$  In accordance with literature



# Methane predictions depending on lactation stage

Application of CH<sub>4</sub> equations on Belgian spectral database



#### Trends over lactations correspond to what is expected











| Institution    | Reference<br>Method      | Number of animals | Number of data |  |
|----------------|--------------------------|-------------------|----------------|--|
| CRA-W          | SF <sub>6</sub>          | 47                | 265            |  |
| Teagasc        | SF <sub>6</sub>          | 110               | 262            |  |
| AFBI           | Chambers                 | 12                | 24             |  |
| Aarhus         | Chambers                 | 19                | 130            |  |
| Qualitas/ETH Z | Chambers/SF <sub>6</sub> | 42 + 16           | 99 + 59        |  |
| FBN            | Chambers                 | 52                | 213            |  |
| Inra           | Chambers                 | 9                 | 82             |  |
| Total          |                          | 307               | 1134           |  |







allonie

RA-W

cherche





















| Equation (g/day)  | Ν    | SD    | R²c  | R <sup>2</sup> cv | SEC | SECV |
|-------------------|------|-------|------|-------------------|-----|------|
| SF <sub>6</sub>   | 446  | 127.5 | 0.75 | 0.67              | 63  | 72   |
| $SF_6$ + Chambers | 1134 | 106.1 | 0.64 | 0.60              | 64  | 67   |

N = number of observations; SD = standard deviation;  $R^2c$  = calibration coefficient of determination;  $R^2cv$  = cross-validation coefficient of determination; SEC = calibration standard error; SECV = cross-validation standard error



#### **Phenotypic and Genetic Large-Scale Studies**





#### **Phenotypic and Genetic Large-Scale Studies**



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#### Genetic parameters of mid-infrared methane predictions and their relationships with milk production traits in Holstein cattle

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#### **Conclusions**

- It is possible to predict methane from milk MIR spectra
- Important to check if the applications at large scale are logical at a metabolic level
- Integration of DIM information seems to be a good strategy to :
  - take a better account of the metabolic status of cows
  - improve the equation
- Important to include further regions/breeds/regimes to cover the variability
- Merging of data set strategy : analytical standardisation of reference measurements is needed
- Easy and cheap method for large scale utilisation



#### Thanks to our support and partners !



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