



# Re-Livestock

RESILIENT FARMING SYSTEMS

## Definition of Methane Phenotypes in Cattle

*Oscar González-Recio and Coralía Manzanilla-Pech*



WAGENINGEN  
UNIVERSITY & RESEARCH



Zaragoza, February 25<sup>th</sup>, 2025

# Data from GreenFeed and C-Locks

Lisanne Koning

30' presentation

10' discussion



# Overview

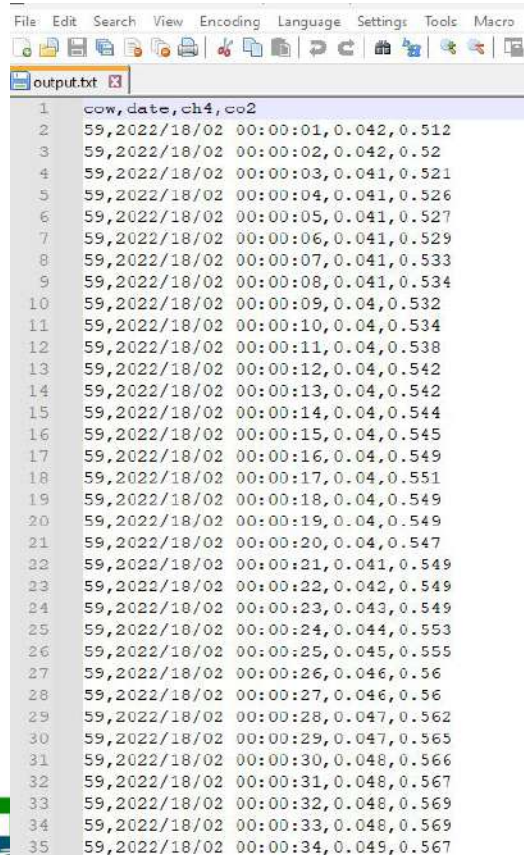
1. Introduction
2. Background/baseline approaches
3. Methane concentration phenotypes (ppm)
4. Methane production phenotypes (g/d)
5. Other methane phenotypes (residuals and ratios)
6. Pros and cons methane phenotypes
7. Overlook of entire process **Netherlands** and **Spain** examples

# Introduction

- Sniffers are used widely to measure  $\text{CH}_4$  for genetic purposes
- Lack of harmonization in the process
- Suggested phenotypes for  $\text{CH}_4$
- or convert it to  $\text{CH}_4$  g/d
- or use another derived phenotype



# Raw CH4 data from sniffer



```
1 cow,date,ch4,co2
2 59,2022/18/02 00:00:01,0.042,0.512
3 59,2022/18/02 00:00:02,0.042,0.52
4 59,2022/18/02 00:00:03,0.041,0.521
5 59,2022/18/02 00:00:04,0.041,0.526
6 59,2022/18/02 00:00:05,0.041,0.527
7 59,2022/18/02 00:00:06,0.041,0.529
8 59,2022/18/02 00:00:07,0.041,0.533
9 59,2022/18/02 00:00:08,0.041,0.534
10 59,2022/18/02 00:00:09,0.04,0.532
11 59,2022/18/02 00:00:10,0.04,0.534
12 59,2022/18/02 00:00:11,0.04,0.538
13 59,2022/18/02 00:00:12,0.04,0.542
14 59,2022/18/02 00:00:13,0.04,0.542
15 59,2022/18/02 00:00:14,0.04,0.544
16 59,2022/18/02 00:00:15,0.04,0.545
17 59,2022/18/02 00:00:16,0.04,0.549
18 59,2022/18/02 00:00:17,0.04,0.551
19 59,2022/18/02 00:00:18,0.04,0.549
20 59,2022/18/02 00:00:19,0.04,0.549
21 59,2022/18/02 00:00:20,0.04,0.547
22 59,2022/18/02 00:00:21,0.041,0.549
23 59,2022/18/02 00:00:22,0.042,0.549
24 59,2022/18/02 00:00:23,0.043,0.549
25 59,2022/18/02 00:00:24,0.044,0.553
26 59,2022/18/02 00:00:25,0.045,0.555
27 59,2022/18/02 00:00:26,0.046,0.56
28 59,2022/18/02 00:00:27,0.046,0.56
29 59,2022/18/02 00:00:28,0.047,0.562
30 59,2022/18/02 00:00:29,0.047,0.565
31 59,2022/18/02 00:00:30,0.048,0.566
32 59,2022/18/02 00:00:31,0.048,0.567
33 59,2022/18/02 00:00:32,0.048,0.569
34 59,2022/18/02 00:00:33,0.048,0.569
35 59,2022/18/02 00:00:34,0.049,0.567
```

# Background approaches

1. Within the visit (milking period)
  - a. Average of the 3-5 lowest values (Spain, Italy, Re-Livestock)
  - b. 0.001 quantile (NL; van Breukelen, et al. 2022) [10.3168/jds.2021-21420](https://doi.org/10.3168/jds.2021-21420)
  - c. median, mode.
2. Outside the visit (non-milking periods)
  - a. using plateau (Poland)
  - b. using a function with sin and cos (Lovendahl, et al. 2024)

<https://doi.org/10.1016/j.compag.2024.109559>

# Background approaches

1. Within the visit (milking period)
  - a. Average of the 3-5 lowest values (Spain, Italy, Re-Livestock)
  - b. 0.001 quantile (NL; van Breukelen, et al. 2022) [10.3168/jds.2021-21420](https://doi.org/10.3168/jds.2021-21420)
  - c. median, mode.
2. Outside the visit (non-milking periods)
  - a. using plateau (Poland, Marcin R.)
  - b. using a function with sin and cos (Lovendahl, et al. 2024)

<https://doi.org/10.1016/j.compag.2024.109559>

# Plateau approach

Methane records in **non-milking** time are divided in **plateau regions**

**Thresholds** for defining a plateau region can be:

Difference between two consecutive records

Min and max methane values

Number of records in a plateau

Finding the **most stable plateaus** depending on the **quantile values** in each plateau

First quantile of  $\text{CH}_4$  value for each **plateau** is the **background  $\text{CH}_4$  value**

**Background** is subtracted from the  $\text{CH}_4$  values in the preceding milking events

# Lovendahl approach

Methane values in **idle periods** should be screened out for outliers based on moving averages of  $\text{CH}_4$  values at each time point

**Threshold** is set for the duration of each **non-milking event**

Records in **initial** and **final** parts of each **non-milking event** are **discarded**

Methane records are analysed using a **linear mixed model**

**Background** values can be **predicted** using the linear model for **each milking event**

Shariff's work



# Diurnal and nocturnal adjustments

1. Daily phenotype → Needed

- Using fourier series (Lovendahl et al. 2006, Lassen et al. 2012)

[https://doi.org/10.3168/jds.S0022-0302\(06\)72404-3](https://doi.org/10.3168/jds.S0022-0302(06)72404-3); <https://doi.org/10.3168/jds.2011-4544>

- can be done in ASReml

2. Weekly phenotype → No needed

# Methane phenotypes

Methane concentration

(MeC)

CH<sub>4</sub> ppm

Methane production

(MeP)

g/d

Residuals and Ratios

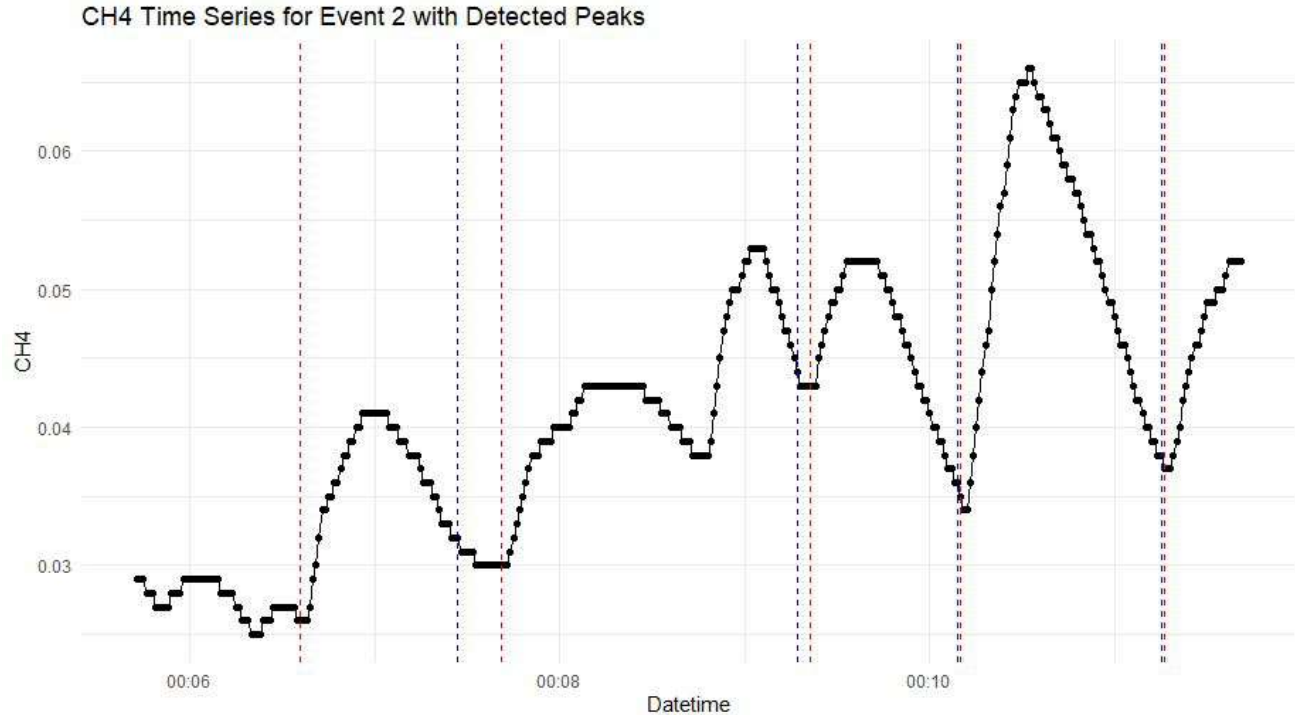
RMet, MeI, MeY

# Phenotypes for CH<sub>4</sub> concentration

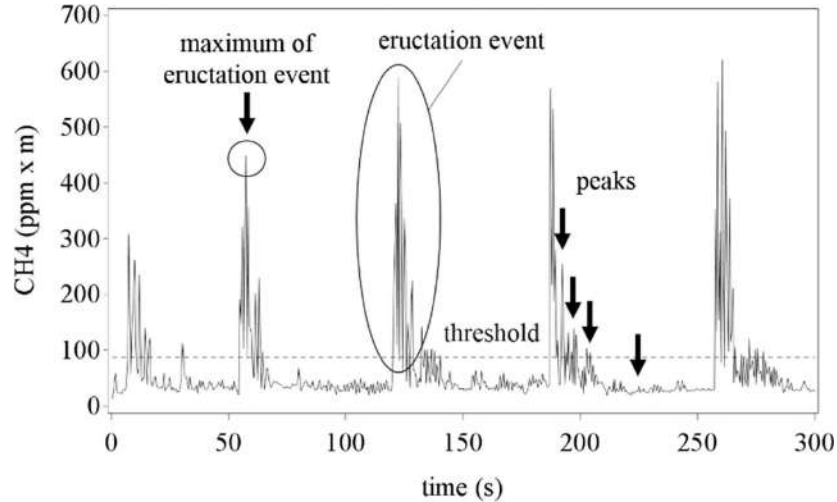
- average methane
  - per visit or minute
  - moving average (dif window size)
- peaks (eructation events)
  - sum/average of max 2 values peaks
  - sum of max peaks
  - number of peaks
- area under the curve
- ratio

# Raw CH4 data from sniffer (Spain)

File	Edit	Search	View	Encoding	Language	Settings	Tools	Macro
output.txt								
1	cow,date,ch4,co2							
2	59,2022/18/02 00:00:01,0.042,0.512							
3	59,2022/18/02 00:00:02,0.042,0.52							
4	59,2022/18/02 00:00:03,0.041,0.521							
5	59,2022/18/02 00:00:04,0.041,0.526							
6	59,2022/18/02 00:00:05,0.041,0.527							
7	59,2022/18/02 00:00:06,0.041,0.529							
8	59,2022/18/02 00:00:07,0.041,0.533							
9	59,2022/18/02 00:00:08,0.041,0.534							
10	59,2022/18/02 00:00:09,0.04,0.532							
11	59,2022/18/02 00:00:10,0.04,0.534							
12	59,2022/18/02 00:00:11,0.04,0.538							
13	59,2022/18/02 00:00:12,0.04,0.542							
14	59,2022/18/02 00:00:13,0.04,0.542							
15	59,2022/18/02 00:00:14,0.04,0.544							
16	59,2022/18/02 00:00:15,0.04,0.545							
17	59,2022/18/02 00:00:16,0.04,0.549							
18	59,2022/18/02 00:00:17,0.04,0.551							
19	59,2022/18/02 00:00:18,0.04,0.549							
20	59,2022/18/02 00:00:19,0.04,0.549							
21	59,2022/18/02 00:00:20,0.04,0.547							
22	59,2022/18/02 00:00:21,0.041,0.549							
23	59,2022/18/02 00:00:22,0.042,0.549							
24	59,2022/18/02 00:00:23,0.043,0.549							
25	59,2022/18/02 00:00:24,0.044,0.553							
26	59,2022/18/02 00:00:25,0.045,0.555							
27	59,2022/18/02 00:00:26,0.046,0.56							
28	59,2022/18/02 00:00:27,0.046,0.56							
29	59,2022/18/02 00:00:28,0.047,0.562							
30	59,2022/18/02 00:00:29,0.047,0.565							
31	59,2022/18/02 00:00:30,0.048,0.566							
32	59,2022/18/02 00:00:31,0.048,0.567							
33	59,2022/18/02 00:00:32,0.048,0.569							
34	59,2022/18/02 00:00:33,0.048,0.569							
35	59,2022/18/02 00:00:34,0.049,0.567							



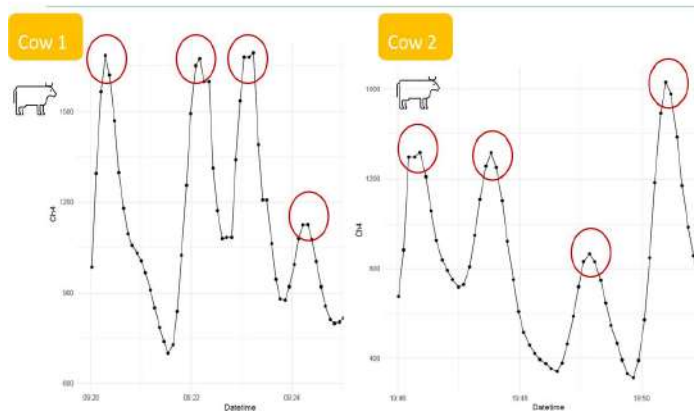
# Methane concentration phenotypes



*Sorg, et al. 2018*



# Methane concentration phenotypes



# Overview of the process



# Editing steps after alignment

Background

- Measuring or estimating



Creating event

- Per animal and visit



Discard first 20-60 sec or sincronice AMS and sniffer events

- Accounting for the time the gas takes to travel through the tube



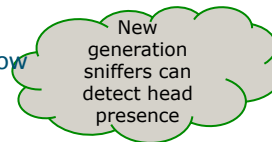
Discard records with less 200 sec

- Keeping homogenous visits



Keep between 60-300 sec

- Head cow in the AMS



Create mean per visit per animal per day

- Delete outliers

**Example (2 months/600 cows)**

**17.6 millions  $\text{CH}_4$ /sec/cow**



**Final  $\text{CH}_4$  conc phenotype**

**24,723 weekly individual  $\text{CH}_4$  records**



# Methane phenotypes

Methane concentration

(MeC)

CH<sub>4</sub> ppm

Methane production

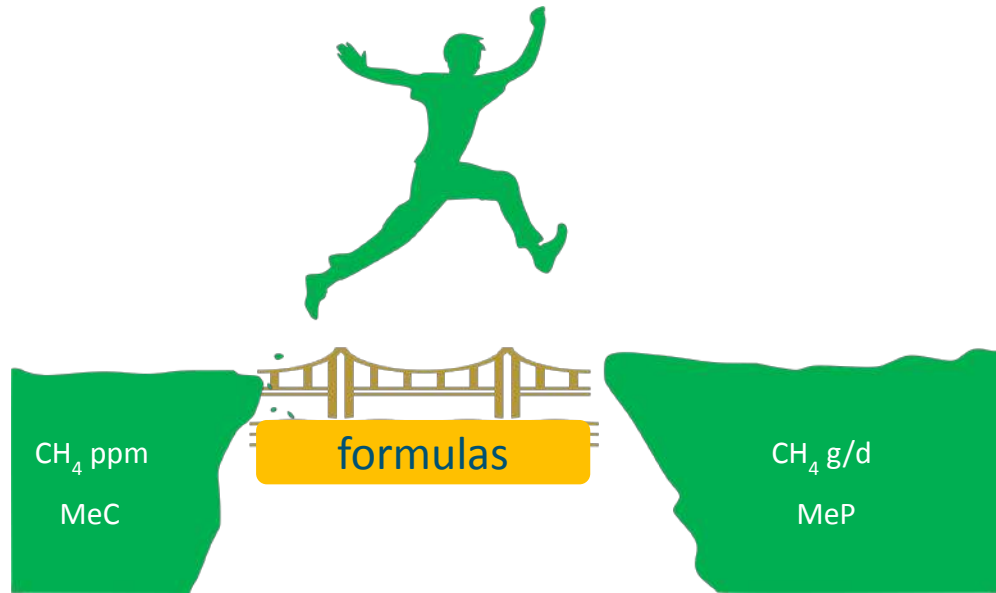
(MeP)

g/d

Residuals and Ratios

RMet, MeI, MeY

# From Methane Concentration to Methane Production



# Methane Production Phenotypes

- With sniffer input
  - Using Ratio: between  $\text{CH}_4$  and  $\text{CO}_2$  concentrations
    - Madsen, Pedersen, Kjeldsen
  - Using  $\text{CH}_4$  concentration
    - Chagunda
- Without sniffer input
  - Based only on DMI, BW and ECM
    - Tier formulas (1,2,3)

# Formulas that predict CO<sub>2</sub>

- Madsen *et al.* (2010) equation <https://doi.org/10.1016/j.livsci.2010.01.001>

CO<sub>2</sub> g/d =  $180 \times 24 \times (5.6 \text{ MBW} + 22 \text{ ECM} + 1.6 \times 10^{-5} \times \text{num days in gestation})$

(BMilkCF, parity  $\times$  Milk CF)

\* CH<sub>4</sub> g/d = ratio  $\times$  predicted CO<sub>2</sub>

# Formulas that predict CO<sub>2</sub>

Kjeldsen et al. 2024 [10.3168/jds.2023-24414](https://doi.org/10.3168/jds.2023-24414)

\* CH<sub>4</sub> g/d = ratio x predicted CO<sub>2</sub>

- Kjeldsen equation 1

$$\text{CO}_2 \text{ (g/d)} = \beta_0 + (\beta_1 \times \text{DMI}) + (\beta_2 \times \text{MBW}) + (\beta_3 \times \text{Diet CP}) + \text{breed} + (\beta_{\text{DMI}}, \text{breed} \times \text{DMI}) + (\beta_{\text{DMI}}, \text{parity} \times \text{DMI}) + (\text{MBW}, \text{breed} \times \text{MBW})$$

- Kjeldsen equation 2

$$\text{CO}_2 \text{ g/d} = \beta_0 + (\beta_1 \times \text{ECM}) + (\beta_2 \times \text{MBW}) + (\beta_3 \times \text{Milk CF}) + (\beta_4 \times \text{DIM}) + \text{breed} + (\beta_{\text{DIM}}, \text{Diet CF} \times \text{DIM} \times \text{Diet CF}) + (\beta_{\text{ECM}}, \text{DIM} \times \text{ECM} \times \text{DIM}) + (\beta_{\text{ECM}}, \text{MBW} \times \text{ECM} \times \text{MBW}) + (\beta_{\text{MilkCF}}, \text{MBW} \times \text{Milk CF} \times \text{MBW}) + (\beta_{\text{MBW}}, \text{breed} \times \text{MBW}) + (\beta_{\text{DIM}}, \text{breed} \times \text{DIM}) + (\beta_{\text{MBW}}, \text{parity} \times \text{MBW})$$

- Kjeldsen equation 3

$$\text{CO}_2 \text{ g/d} = \beta_0 + (\beta_1 \times \text{ECM}) + (\beta_2 \times \text{DIM}) + \text{breed} + \text{parity} + (\beta_{\text{breed}}, \text{parity}) + (\beta_{\text{DIM}}, \text{Diet CF} \times \text{DIM} \times \text{Diet CF}) + (\beta_{\text{ECM}}, \text{DIM} \times \text{ECM} \times \text{DIM}) + (\beta_{\text{DIM}}, \text{breed} \times \text{DIM}) + (\beta_{\text{MilkCF}}, \text{parity} \times \text{Milk CF})$$

Item	Model 1	Model 2	Model 3
Intercept	956	-6,134	8,781
DMI (kg/d)	122		
ECM (kg/d)		213	80.3
MetaBW (kg)	60.4	126	
Diet CP (g/kg DM)	3.44		
Milk CF (g/kg)		52.5	
DIM (d)		-5.13	-4.66
Breed			
Ayrshire	0	0	0
Holstein	-777	2,117	-49.0
Jersey	1,103	1,364	-2,321
Others/crossbreeds	1,501	4,083	-1,237
Parity			
First			0
Second			511
Third and higher			1,587
DIM × Diet CF		-0.122	-0.149
ECM × DIM		0.386	0.338
ECM × metaBW		-1.18	
Milk CF × metaBW		-0.614	
DMI × Ayrshire	0		
DMI × Holstein	206		
DMI × Jersey	204		
DMI × others/crossbreeds	225		
DMI × first parity	0		
DMI × second parity	7.53		
DMI × third parity	15.7		
MetaBW × Ayrshire	0	0	
MetaBW × Holstein	-18.5	-5.96	
MetaBW × Jersey	-37.3	-1.03	
MetaBW × others/crossbreeds	-43.2	-33.4	
DIM × Ayrshire		0	0
DIM × Holstein		2.06	6.05
DIM × Jersey		2.49	6.02
DIM × others/crossbreeds		8.94	11.3
MetaBW × first parity		0	
MetaBW × second parity		3.66	
MetaBW × third parity		4.01	
First parity × milk CF			-4.18
Second parity × milk CF			-10.5
Third parity × milk CF			-28.8
Ayrshire × first parity			0
Ayrshire × second parity			0
Ayrshire × third parity			0
Holstein × first parity			0
Holstein × second parity			775
Holstein × third parity			803
Jersey × first parity			0
Jersey × second parity			608
Jersey × third parity			1,307
Others/crossbreeds × first parity			0
Others/crossbreeds × second parity			791
Others/crossbreeds × third parity			659

## Coefficients for the different models



<sup>1</sup>Diet CF = dietary crude fat (g/kg DM), diet CP = dietary crude protein (g/kg DM), DIM = days in milk (d), DMI = dry matter intake (kg/d), ECM = energy-corrected milk yield (kg/d), milk CF = milk crude fat (g/

# Formulas that predict CH<sub>4</sub>

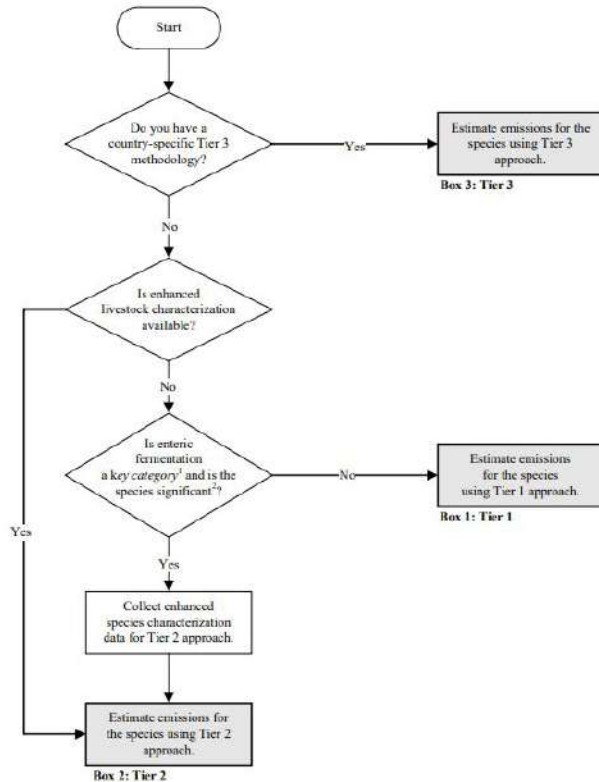
- Chagunda *et al.* (2009) equation <https://doi.org/10.1016/j.compag.2009.05.008>

$$\text{CH}_4 \text{ g/d} = 0.000576 \times M_{\text{TV}} \times \text{TV}_r$$

$$M_{\text{TV}} = \text{CH}_4 \text{ conc} \times \text{TV}_r$$

# Formulas that predict CH<sub>4</sub>: TIERs

Figure 10.2 Decision Tree for CH<sub>4</sub> Emissions from Enteric Fermentation



IPCC, 2019

[https://www.ipcc-nggip.iges.or.jp/public/2006gl/pdf/4\\_Volume4/V4\\_1\\_0\\_Ch10\\_Livestock.pdf](https://www.ipcc-nggip.iges.or.jp/public/2006gl/pdf/4_Volume4/V4_1_0_Ch10_Livestock.pdf)

**EQUATION 10.21**  
**CH<sub>4</sub> EMISSION FACTORS FOR ENTERIC FERMENTATION FROM A LIVESTOCK CATEGORY**

$$EF = \left[ \frac{GE \cdot \left( \frac{Y_m}{100} \right) \cdot 365}{55.65} \right]$$

where:

EF = emission factor, kg CH<sub>4</sub> head<sup>-1</sup> yr<sup>-1</sup>

GE = gross energy intake, MJ head<sup>-1</sup> day<sup>-1</sup>

Y<sub>m</sub> = methane conversion factor, per cent of gross energy in feed converted to methane

The factor 55.65 (MJ/kg CH<sub>4</sub>) is the energy content of methane



**EQUATION 10.16**  
**GROSS ENERGY FOR CATTLE/BUFFALO AND SHEEP**

$$GE = \left[ \frac{\left( \frac{NE_m + NE_a + NE_l + NE_{work} + NE_p}{REM} \right) + \left( \frac{NE_g + NE_{wool}}{REG} \right)}{\frac{DE\%}{100}} \right]$$

ere:

GE = gross energy, MJ day<sup>-1</sup>

NE<sub>m</sub> = net energy required by the animal for maintenance (Equation 10.3), MJ day<sup>-1</sup>

NE<sub>a</sub> = net energy for animal activity (Equations 10.4 and 10.5), MJ day<sup>-1</sup>

NE<sub>l</sub> = net energy for lactation (Equations 10.8, 10.9, and 10.10), MJ day<sup>-1</sup>

NE<sub>work</sub> = net energy for work (Equation 10.11), MJ day<sup>-1</sup>

NE<sub>p</sub> = net energy required for pregnancy (Equation 10.13), MJ day<sup>-1</sup>

REM = ratio of net energy available in a diet for maintenance to digestible energy consumed (Equation 10.14)

NE<sub>g</sub> = net energy needed for growth (Equations 10.6 and 10.7), MJ day<sup>-1</sup>

NE<sub>wool</sub> = net energy required to produce a year of wool (Equation 10.12), MJ day<sup>-1</sup>

REG = ratio of net energy available for growth in a diet to digestible energy consumed (Equation 10.15)

DE% = digestible energy expressed as a percentage of gross energy

## MeP

- ✓ Easy to merge with other countries that use other methods
- ✓ Easy to explain and compare with other traits with the same unit
- ✗ Total dependency on ECM and BW
- ✗ Problems with double counting

## MeC

- ✓ No induced correlation with ECM and BW
- ✗ Difficult to compare with other traits
- ✗ Concentration does not account for size and production of the animal

# Methane phenotypes

Methane concentration

(MeC)

CH<sub>4</sub> ppm

Methane production

(MeP)

g/d

Residuals and Ratios

RMet, MeI, MeY

# Other Methane Phenotypes

## RESIDUAL

RMeP

1. Regression on MBW and DMI and fixed effects
2. Regression on MBW and ECM and fixed effects

RMeC

3. Regression on MBW, DMI and ECM and fixed effects

## RATIO

Methane intensity (MeI)

$\text{g CH}_4 / \text{kg ECM}$

Methane yield (MeY)

$\text{g CH}_4 / \text{kg DMI}$   
 $\text{g CH}_4 / \text{kg BW}$

## Residual traits

- ✓ It is independent of ECM and BW
- ✓ Easy to rank animals
- ✓ Could be use with MeC to account for ECM and BW
- ✓ Easy and effective to include in the breeding goal
- ✗ Can be seen as an index inside an index
- ✗ Not easy to explain to farmers

## Ratio traits

- ✓ Nutritionist and farmers prefer it
- ✓ Easy to explain in terms of g  $\text{CH}_4$  per kg of milk, feed or body weight
- ✗ Dependency with the numerator
- ✗ Correlation with ECM, DMI, or BW

Calculation of baseline and merging  
data with milk recording: The  
Netherlands example



# Downloading data from Azure database



Sniffers (Carltech BV):

- WD-WUR v1.0 + Arduino (KE)
- WD-WUR v2.0 (PPS CSCB)

## Recording

Anouk's work

Connect sniffer to database with  
MethaanWatcher azurewebsite

Twice a week visually check if  
the sniffers work correctly with  
wlrsniffer azurewebsite  
(wlrsniffer.azurewebsites.net)

Microsoft Azure database:  
euwstkemethanadatap -> container newdata  
(csv file per farm per day with CH<sub>4</sub>, CO<sub>2</sub>, and  
time of day)

Milkingrobot data from CRV (cow ID and date and time of  
milking) and cow information (Parity, Breed, calving date)

Python script to download the data to a local computer or  
HPC, filter out bad data, and connect to cow ID

Python script to prepare data for genetic  
evaluation by adding cow information

# Data processing 1

Pipeline\_part1.sh, runs scripts:

Python Script number	Function
1	Makes a subfile of the AMS data per farm, to increase computational efficiency
2	Make a list of all the farms that are in the Azure database with sniffer data, to download them one by one for each farm
3	Processes the .csv files in the Azure database one by one, and calls functions to filter the data, and to align the filtered data to the AMS data



# Data processing 2

Pipeline\_part2.sh, runs scripts:

Script number	Function
4	Calculates the mean emissions per milking robot visit, including some filtering criteria of which records to include in a visit
5	This script combines the methane data with data on other cow traits for genetic analyses. Based on the animals in the combined dataset the pedigree is then pruned

# Data processing 3

Run\_diu.sh, runs scripts:

Script name	Function
1.R	Runs ASReml to precorrect visit means for diurnal variation, before averaging to daily means. The model includes a random genetic and permanent environmental effect

# Data processing 4

Pipeline\_part3.sh, runs scripts:

Script name	Function
6.py	Add the diurnal corrected trait from the ASReml output, and summarize traits as weekly means

# Data filtering criteria for CH<sub>4</sub> – Raw sniffer data

- Groups data by farm, date and hour, NA if:
  - Interquartile larger than 200
  - Less than 30% of data should be within 10 ppm (2 first modes)
  - Not measurements above 3500 ppm
- Individual measurements below the 0.001 quantile and above the 0.999 quantile = NA
- Within-day scaled and centred phenotype is made
- Data is matched to the AMS data, AMS times are shifted by one minute because of the delay in when concentrations reach the sniffer

# Data filtering criteria for CH<sub>4</sub> – Visit means

- Background concentrations as the 0.001 lowest quantile per farm per day
- Summarises the data and calculate means by milking robot visit
- Keep only records between 60 and 300 seconds of milking
- Remove records of less than 150 seconds during milking
- A file used for diurnal correction in ASReml is made, including:
  - Parity and calf dates (not necessary anymore when not analysing visits, may remove later)

# Data filtering criteria – Genetic analyses

- Adds cow information (e.g. calving info) and other traits
- Prune the pedigree
- After running ASReml the fixed effect solutions are subtracted from the  $\text{CH}_4$  trait
- Week records with less than 7  $\text{CH}_4$  records = NA
- Records = means per calendar week
- Keep records up to 405 DIM
- Discard records of cows < 25% HF
- Data is standardized within project (KE or PPS)
- Discard lactations with less than 3 weekly records

# Methane Team at ABG WUR

Present at this course

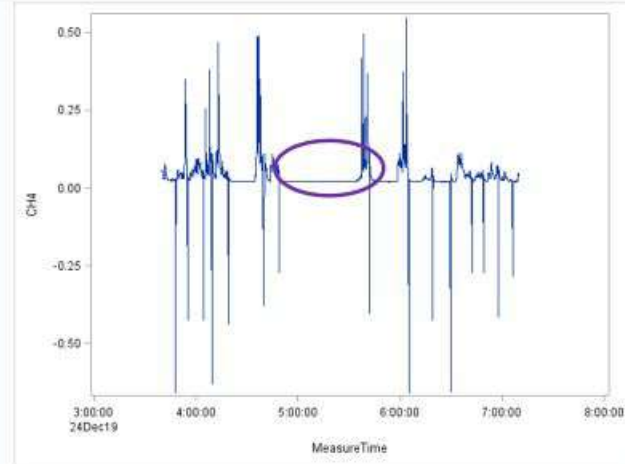
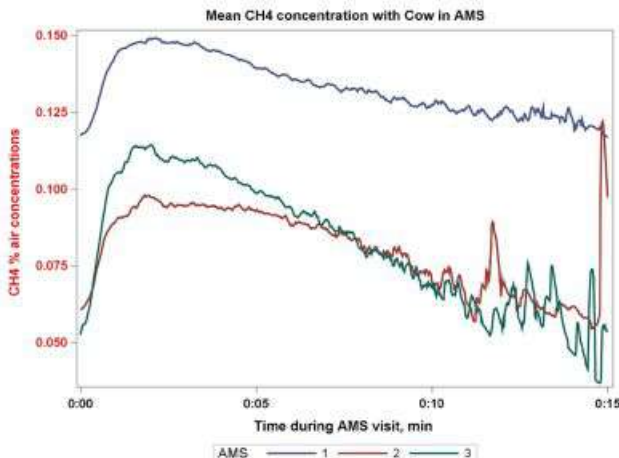


Calculation of baseline and  
merging data with milk recording:  
The **Denmark** example



# Aligning the data and setting a background

Variation in CH<sub>4</sub> conc across minutes



Identify cleaning time of AMS



# Aligning the data (Developed software)



Computers and Electronics in Agriculture

Volume 201, October 2022, 107299



Data synchronization for gas emission measurements from dairy cattle: A matched filter approach

Viktor Milkevych <sup>a</sup>, Trine Michelle Villumsen <sup>a</sup>, Peter Lavendahl <sup>a</sup>, Goutam Sahana <sup>b</sup>



Computers and Electronics in Agriculture

Volume 227, Part 1, December 2024, 109559



A data-driven approach to the processing of sniffer-based gas emissions data from dairy cattle

Peter Lavendahl <sup>a</sup>, Viktor Milkevych <sup>a</sup>, Rikke Krogh Nielsen <sup>a</sup>, Martin Bjerring <sup>b</sup>,  
Coralia Manzanilla-Pech <sup>a</sup>, Kresten Johansen <sup>a</sup>, Gareth F Difford <sup>a</sup>, Trine M Villumsen <sup>a</sup>

On-site Automated Milking System



On-site Sniffers System



AMS database



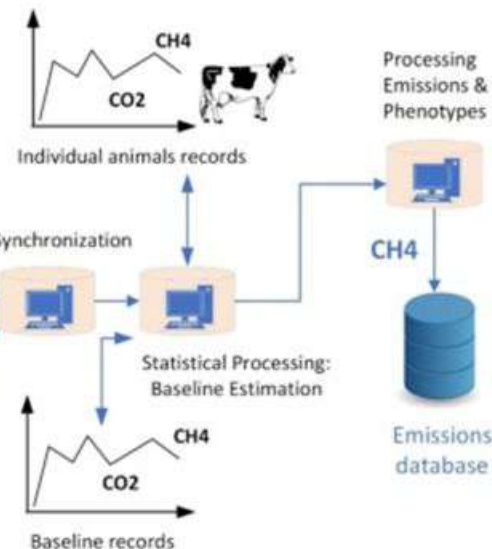
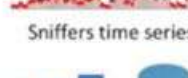
AMS time series



Sniffers database



Sniffers time series

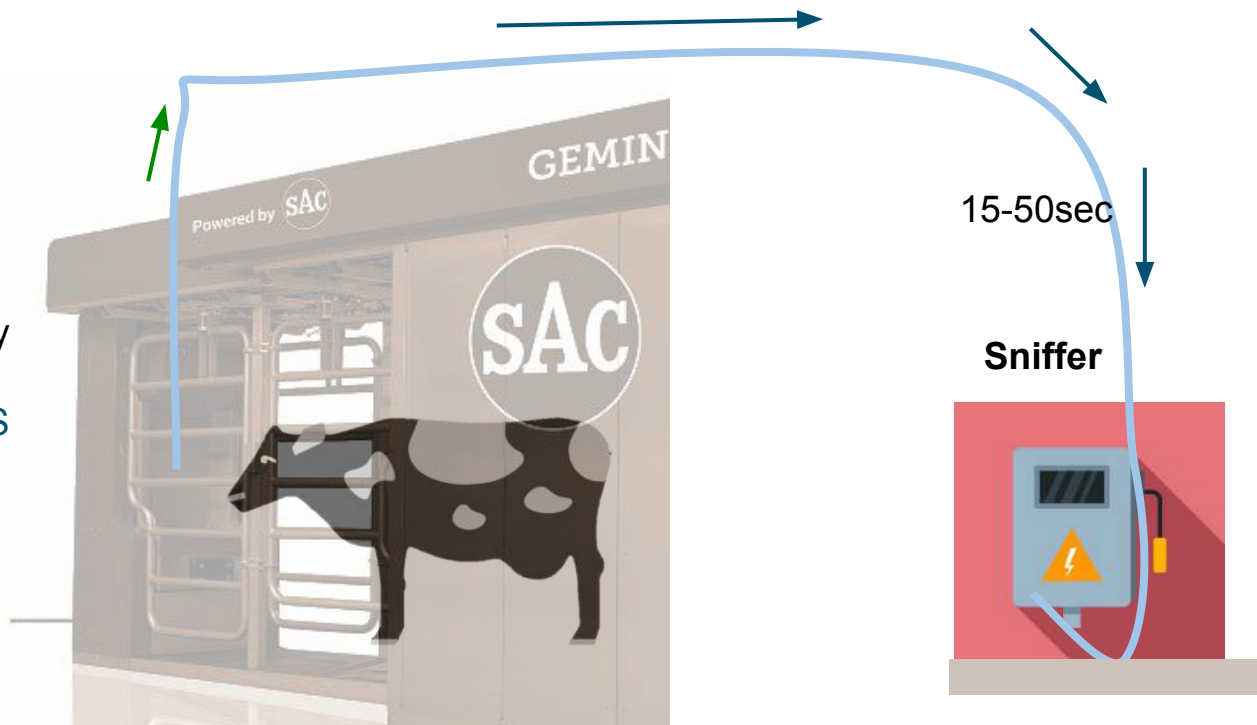


Calculation of baseline and  
merging data with milk recording:  
Spain example



# Installation and lag calculation

- Lag: Calculate the time that the eructation takes to arrive in the sensor.
- Longer tubes, longer lag.
- Tube obturation may change the lag period. Monitorize daily/weekly
- Synchronize clocks from the AMS and Sniffer, and take into account the lag to assign the event to the right cow (*in-house software*).





## Recording

Lag is calculated  
Download data files from sniffer  
and AMS on site.

Combine files.

# Raw Data



Sincronize



SUAREZ\_102

SANCHEZdic21\_101\_Loggy\_42

Date and time	CH4	CO2
03/12/2021 11:50:00	0,018	0,095
03/12/2021 11:50:01	0,018	0,095
03/12/2021 11:50:02	0,018	0,093
03/12/2021 11:50:03	0,018	0,094
03/12/2021 11:50:04	0,018	0,093
03/12/2021 11:50:05	0,018	0,095
03/12/2021 11:50:06	0,018	0,093
03/12/2021 11:50:07	0,018	0,095
03/12/2021 11:50:08	0,018	0,095
03/12/2021 11:50:09	0,018	0,095
03/12/2021 11:50:10	0,018	0,095
03/12/2021 11:50:11	0,018	0,095
03/12/2021 11:50:12	0,019	0,095
03/12/2021 11:50:13	0,019	0,095
03/12/2021 11:50:14	0,019	0,094
03/12/2021 11:50:15	0,019	0,093
03/12/2021 11:50:16	0,019	0,094
03/12/2021 11:50:17	0,02	0,095
03/12/2021 11:50:18	0,02	0,095
03/12/2021 11:50:19	0,02	0,095
03/12/2021 11:50:20	0,02	0,095
03/12/2021 11:50:21	0,02	0,095
03/12/2021 11:50:22	0,02	0,097
03/12/2021 11:50:23	0,02	0,096
03/12/2021 11:50:24	0,021	0,095
03/12/2021 11:50:25	0,021	0,095

Numero vaca	Address	Fecha/Hora de visita	tiempo en cubiculo	Produccion de leche	Tiempo	Descripcion	CIB
161	102	22/11/2021 0:02:00	4:26	18.8	6:04	Correcto	ES031112194660
148	102	22/11/2021 0:15:00	10:08	10.9	11:39	Correcto	ES021112624231
193	102	22/11/2021 0:23:00	6:01	15.8	7:38	Correcto	ES091111967021
110	102	22/11/2021 0:28:00	3:45	14.4	5:04	Correcto	ES041112449478
136	102	22/11/2021 0:35:00	5:37	15.1	7:11	Correcto	ES091112194713
158	102	22/11/2021 0:43:00	5:42	15.9	7:29	Correcto	ES031112449535
257	102	22/11/2021 0:48:00	3:41	11.3	5:20	Correcto	ES091112624238
178	102	22/11/2021 0:54:00	3:44	14.6	5:24	Correcto	ES061112449527
142	102	22/11/2021 1:01:00	5:01		15 6:35	Correcto	ES031112194693



CIB	cow	date	hour	time	meanCH4	meanCO2	meanRatioCH4 CO2	peaks	Sum of PeaksCH4	Sum of PeaksCO2	Sum of PeaksRatio	AUC CH4	AUC CO2	AUC Ratio	Mean of PeaksCH4	Mean of PeaksCO2	Mean of PeaksRatio	validity
ES01111344001	210	Mon Nov 22 19:48:35 CET 2021	19:1	180.0	388.52	522.8	0.28	8	1482.80711845036	3404.7505912292	0.289055540186731	9078.99	1.37	247.134519100867	0.0304845207334455	0		0
ES01111344001	210	Mon Nov 22 19:50:33 CET 2021	6:7	218.0	343.03	459.27	0.07	3	362.1901794365187	1536.941090910792	0.14045011982617295	4676.82	1.8	130.732264730329	0.0484542466214315	11		11
ES021112449404	196	Mon Nov 22 20:05:54 CET 2021	15:7	257.0	77.83	3027.08	0.07	3	227.29635845411727	11962.701820728262	0.125719611820071	2248.25	1.19	75.7051028780579	0.041005537200693	10		10
ES041111966905	275	Mon Nov 22 20:13:18 CET 2021	12:9	301.0	158.91	3602.40	0.11	5	485.0025080060604	16271.32819301367	0.120701320773351	3745.81	4.02	97.0010364133034	0.0161014085403062	10		10
ES01111344001	210	Mon Nov 22 20:19:49 CET 2021	16:4	302.0	383.1	5693.08	0.09	5	943.700180000722	35488.2173013675	0.1171030382020392	6333.54	0.8	188.7410398001446	0.02342016234007014	0		0
ES01111344001	210	Mon Nov 22 20:25:13 CET 2021	6:2	418.0	173.88	4023.8	0.09	8	595.89141891545	33887.90164801919	0.2010510134040027	5467.48	1.23	188.116322350326	0.0303403866514526	0		0
ES01111344001	210	Mon Nov 22 20:34:27 CET 2021	14:2	284.0	18.84	151.71	0.13	1	136.775781700104	2414.15103032019768	0.1070598617604006	1470.42	4.0	38.7521036861746	0.1070598617604006	10		10
ES011121947021	209	Mon Nov 22 20:40:08 CET 2021	8:8	312.0	125.4	310.7	0.09	1	18333.22326	15736.301000204608	0.1794067802104208	4218.05	1.7	113.22080000081345	0.04469119007080310	1		1
ES01112448016	168	Mon Nov 22 20:48:18 CET 2021	20:8	353.0	182.3	364.31	0.07	5	13623.00000	18031.8000524136	0.152432880784817	4350.2	2.36	115.59447200212525	0.0630586873408633	10		10
ES011121947021	209	Mon Nov 22 20:50:29 CET 2021	14:0	340.0	64.00	140.72	0.12	4	15000.74001	19528.440000000002	0.4200130200033406	2048.12	3.33	65.44510794727794	0.1103037171208064	0		0
ES01112448016	168	Mon Nov 22 21:06:18 CET 2021	11:7	353.0	124.63	205.61	0.12	3	416.18450100100	11718.4140000323	0.14671027200114096	2716.07	1.12	128.732068328796	0.050706969711684	10		10
ES01111344001	210	Mon Nov 22 21:35:45 CET 2021	18:4	386.0	343.62	3627.08	0.1	5	638.158128005367	20341.160161028847	0.3886115102411604	11601.1	3.96	127.617747610863	0.1760232510305295	10		10
ES01112448016	168	Mon Nov 22 21:47:08 CET 2021	12:1	180.0	107.48	2412.15	0.23	7	1007.60180400406	21018.304000300356	0.330211447947989	4626.87	0.86	172.5140980414805	0.118293747080227	0		0
ES01112448016	168	Mon Nov 22 21:54:12 CET 2021	12:3	238.0	62.29	3033.13	0.11	2	211.11220611220611	9635.07184021848	0.08837873015197078	3180.75	1.63	105.55132005133009	0.0341930757588329	10		10
ES011119807021	223	Mon Nov 22 22:03:30 CET 2021	19:4	409.0	65.35	1730.48	0.12	5	348.903172802304	11015.21048000000	0.4280311793201004	3536.85	4.5	88.8817447148468	0.0860751030471921	0		0
ES01111344001	210	Mon Nov 22 22:10:03 CET 2021	21:8	380.0	102.21	624.77	0.06	4	665.80690060607	24487.1012121212	0.2342500016001	8774.83	2.08	214.7421642414245	0.0887016482004028	10		10
ES011119807021	223	Mon Nov 22 22:16:43 CET 2021	11:1	208.0	250.28	678.11	0.06	4	569.2437406500609	20125.21042001548	0.1765406167125346	6377.46	2.02	142.310261015822	0.0438841541781385	10		10
ES011121947021	209	Mon Nov 22 22:22:57 CET 2021	10:8	674.0	140.17	6116.07	0.04	11	1403.320170111370	72146.30100470487	0.280410920335346	1054.48	1.75					
ES01112448016	168	Mon Nov 22 22:36:04 CET 2021	13:2	307.0	128.16	4703.08	0.06	4	376.3674521700728	20420.723081305484	0.104053266331182	3368.29	1.65					
ES01111344001	210	Mon Nov 22 22:43:15 CET 2021	18:3	416.0	103.82	2997.8	0.09	4	434.4544789154204	15196.32139471518	0.236091020813008	4414.04	2.8					
ES011119807021	223	Mon Nov 22 23:05:08 CET 2021	16:4	671.0	124.73	3133.5	0.11	6	676.4545140194419	24473.32107010232	0.146010084720821	4110.94	2.12					



# OUTPUT

SUAREZ\_102output

CIB	cow	date	kgm	time	meanCH4	meanCO2	meanRatioCH4_CO2	peaks	Sum_of_PeaksCH4	Sum_of_PeaksCO2	Sum_of_PeaksRatio	AUC_CH4	AUC_Ratio	Mean_of_PeaksCH4	Mean_of_PeaksRatio	validity
ES041112449525	230	Mon Nov 22 19:48:33 CET 2021	17.1	398.0	388.52	5224.3	0.08	6	1482.8071148459383	36504.70058512293	0.20990055440186733	9978.99	1.37	247.1345191409897	0.03498342573364455	0
ES071112449528	121	Mon Nov 22 19:59:33 CET 2021	0.7	218.0	143.03	4209.27	0.07	3	392.1966794380587	13818.945593869732	0.14540231828572295	4879.82	1.9	130.7322264793529	0.048467439428574316	11
ES021112449545	186	Mon Nov 22 20:06:54 CET 2021	15.7	257.0	77.83	3027.98	0.07	3	227.29635854341737	11562.271820728292	0.1257196611920071	2248.25	1.19	75.76545284780579	0.04190655373066903	10
ES041111966965	275	Mon Nov 22 20:13:10 CET 2021	12.0	301.0	159.91	3662.45	0.11	5	485.00529320866624	18271.33913505767	0.5307052327703151	3715.61	4.02	97.00105864133324	0.10614104655406302	10
ES011112449533	368	Mon Nov 22 20:19:49 CET 2021	10.4	322.0	383.1	5698.58	0.09	5	943.7062980030722	35488.7873015873	0.11710308352325392	6333.54	0.8	188.74125960061446	0.023420616704650784	0
ES061112449505	160	Mon Nov 22 20:25:13 CET 2021	0.0	418.0	173.68	4022.3	0.09	6	996.6991341991343	33887.80194805195	0.20150781394808717	5467.46	1.23	166.1165223665224	0.033584635658014526	0
ES051112449526	30	Mon Nov 22 20:34:27 CET 2021	14.2	284.0	58.84	1541.71	0.13	1	39.7752808988764	1414.1573033707866	0.10767299876464806	1470.42	4.0	39.7752808988764	0.10767299876464806	10
ES091112194724	259	Mon Nov 22 20:40:59 CET 2021	9.6	312.0	125.06	3196.0	0.05	4	452.8353227232538	15736.551503094608	0.17945967803194526	4218.08	1.7	113.20883068081345	0.044864919507986316	1
ES061112449516	168	Mon Nov 22 20:48:18 CET 2021	20.6	353.0	182.54	3645.61	0.07	5	577.9823654606263	18021.85005574136	0.3152943286704917	4330.2	2.38	115.59647309212525	0.06305886573409833	10
ES041112194694	123	Mon Nov 22 20:56:29 CET 2021	14.0	342.0	84.08	1892.72	0.18	4	329.7845117845118	9928.445993265992	0.4332130926953546	2649.12	3.33	82.44612794612794	0.10830327317383864	0
ES081112642404	364	Mon Nov 22 21:25:18 CET 2021	17.7	353.0	124.63	2983.51	0.12	3	416.18426501035196	11718.46149068323	0.16710207260913496	2575.07	1.12	138.728088338784	0.055700690869711654	10
ES071112642447	253	Mon Nov 22 21:33:45 CET 2021	16.4	386.0	143.62	3937.58	0.1	5	638.1588738059327	20841.785816728847	0.38961611552911024	6160.1	3.95	127.63177476118863	0.07792322310582205	10
ES051112642398	179	Mon Nov 22 21:42:08 CET 2021	12.1	603.0	187.48	2412.15	0.23	7	1207.601986249045	21018.304430863256	1.353011447947969	6426.87	9.86	172.51456946414928	0.1932873497068527	0
ES081112642437	153	Mon Nov 22 21:54:12 CET 2021	12.3	239.0	93.29	3023.13	0.11	2	211.11226611226613	9635.074844074843	0.06837873515197078	3180.75	1.03	105.55613305613306	0.03418936757598539	0
ES081111967020	233	Mon Nov 22 22:00:32 CET 2021	12.4	459.0	65.35	1708.46	0.12	5	349.9339738923324	11015.28748938095	0.42803751793536104	3556.86	4.5	69.98679477846648	0.0856070358707221	0
ES011112194715	243	Mon Nov 22 22:10:03 CET 2021	21.5	308.0	232.21	5241.77	0.06	4	866.969696969697	24427.121212121212	0.2342820593185971	8578.83	2.08	216.74242424242425	0.05857051482964928	10
ES011111966973	135	Mon Nov 22 22:16:43 CET 2021	11.1	268.0	220.28	5748.11	0.06	4	569.2437160540609	25125.016420361248	0.17554566167125346	6377.46	2.02	142.31092901351522	0.043886415417813365	10
ES051112194684	180	Mon Nov 22 22:22:57 CET 2021	10.8	674.0	140.17	6116.57	0.04	11	1423.3539729111376	72146.39184785487	0.3804216924355346	6294.48	1.72	129.39581571919433	0.03458379022141223	10
ES091112449542	189	Mon Nov 22 22:36:04 CET 2021	13.2	307.0	128.16	4759.58	0.06	4	376.9674621709726	22420.723981900454	0.10450932553531182	3536.29	1.02	94.24186554274316	0.026127331383827956	10
ES031112449535	158	Mon Nov 22 22:43:15 CET 2021	18.3	416.0	103.82	2997.8	0.09	4	434.45247897842264	15196.337130475158	0.2580819528163808	4414.04	2.94	108.61311974410566	0.0645204882040952	10
ES091111966982	246	Mon Nov 22 22:52:00 CET 2021	16.4	671.0	134.73	3133.5	0.11	6	878.4542245194419	24473.33101922232	0.5466706894738621	4159.94	2.15	146.40903741990698	0.09111178157897702	0

# Raw Data

## OUTPUT



ControlDiciembre2021SUAREZ

SUAREZ_10output																		
CIB	cow	date	kgm	time	meanCH	meanCO2	meanRatioCH_CO2	peaks	Sum of PeaksCH	Sum of PeaksCO2	Sum of PeaksRatio	AUC_CH	AUC_Ratio	Mean of PeaksCH	Mean of PeaksRatio	validity		
ES041112449525	230	Mon Nov 22 19:48:33 CET 2021	17.1	398.0	388.52	5224.3	0.08		6	1482.8071148453383	36504.7058512203	0.2089005540186733	9678.99	1.37	247.1345191408897	0.0348934257354455	0	
ES071112448520	131	Mon Nov 22 19:59:33 CET 2021	0.7	218.0	143.03	4209.27	0.07		3	382.19657942801587	13818.945593891752	0.14542031808672296	4878.82	1.9	130.732294740329	0.048467428426574315	11	
ES021112449545	186	Mon Nov 22 20:05:54 CET 2021	15.7	257.0	77.83	3027.98	0.07		3	227.2963585431737	11562.271820728292	0.128719661192071	2248.25	1.19	75.78552084795579	0.04190855373068903	0	
ES041112449565	275	Mon Nov 22 20:13:10 CET 2021	12.0	301.0	159.91	3662.45	0.11		5	485.00592409604924	18921.3971350527467	0.5307052397703151	3715.61	4.02	97.0015084133324	0.10811414605403032	0	
ES041112449533	368	Mon Nov 22 20:19:49 CET 2021	10.4	322.0	130.3	3693.58	0.09		5	943.7029189003722	30489.21730735172	0.1171030320325382	6333.34	0.8	188.74729900041440	0.020405167649590794	0	
ES061112449505	160	Mon Nov 22 20:25:15 CET 2021	0.0	414.0	173.88	4022.3	0.09		6	996.889141991343	33887.8164801195	0.50750195496717	5487.48	1.23	186.1165229960224	0.026364983656514026	0	
ES051112449520	30	Mon Nov 22 20:34:27 CET 2021	14.3	284.0	58.86	1541.71	0.13		1	38.77530306889704	1414.5173203703968	0.10767979876464006	1470.42	4.0	38.77530306889704	0.10767979876464006	0	
ES091112449424	259	Mon Nov 22 20:40:59 CET 2021	9.6	312.0	126.06	3189.0	0.05		4	422.833227232358	16736.91160201409	0.17945978630194826	4218.08	1.7	113.20583066081435	0.044684919907989319	1	
ES061112449516	169	Mon Nov 22 20:48:18 CET 2021	20.8	353.0	182.54	3645.61	0.07		5	577.9823564606263	18021.80505574136	0.3192493286701847	4330.2	2.38	115.59547200612520	0.0630088573408833	0	
ES081112449404	123	Mon Nov 22 20:50:29 CET 2021	14.0	342.0	84.06	1892.72	0.18		4	329.7845117845118	9508.445592630392	0.43302130102603548	2148.12	3.33	82.44512794617954	0.1083037371735	0	
ES081112449424	264	Mon Nov 22 21:25:18 CET 2021	17.2	353.0	124.63	2083.51	0.12		3	416.18424561035196	11718.46148068323	0.18710207290013496	2578.07	1.32	138.7230868330784	0.08570090909090909	0	
ES071112449447	213	Mon Nov 22 21:33:45 CET 2021	16.4	386.0	143.62	3917.58	0.1		5	638.158973809397	20841.738187278947	0.38861611505911004	6160.1	3.95	127.831774767188651	0.07790255555555555	0	
ES051112449530	179	Mon Nov 22 21:42:08 CET 2021	12.1	603.0	167.48	2412.15	0.13		7	1707.601892440405	20109.30443893256	1.53201447947969	9426.87	5.86	172.0140884414001	0.18305748925555555	0	
ES091112449427	153	Mon Nov 22 21:54:12 CET 2021	12.3	239.0	85.29	3023.13	0.11		2	211.112261122611	9636.501640271043	0.08679178197878	1160.75	1.02	105.59133265133005	0.02419049797594	0	
ES091112449520	233	Mon Nov 22 22:02:32 CET 2021	12.6	409.0	65.36	1708.46	0.12		5	345.9333726902354	11015.0140903305	0.42803717930136104	8506.85	4.5	68.9632947764648	0.0696023263670221	0	
ES011112449476	243	Mon Nov 22 22:10:03 CET 2021	21.8	308.0	232.21	6241.77	0.06		4	866.960696969697	24427.121212121212	0.234382059018971	8578.83	2.08	216.74242424242422	0.08870514820549028	0	
ES011112449573	135	Mon Nov 22 22:16:43 CET 2021	11.1	268.0	220.28	5748.11	0.06		4	589.243710054009	20125.016420912148	0.17954566167125346	6377.46	2.02	129.30181571919433	0.034868115417813365	0	
ES051112449404	180	Mon Nov 22 22:22:57 CET 2021	10.8	674.0	140.17	6115.67	0.04		11	1403.303720911376	72146.38184785487	0.38042161824303546	6294.48	1.72	129.30181571919433	0.034868115417813365	0	
ES081112449542	189	Mon Nov 22 22:36:04 CET 2021	13.2	307.0	126.16	4763.68	0.06		4	376.9874621700726	22420.723081900454	0.10450032553531182	3526.29	1.02	94.24185564274318	0.026127331383527955	0	
ES031112449533	138	Mon Nov 22 22:43:15 CET 2021	18.3	416.0	103.82	2991.8	0.09		4	434.45247879542564	10196.337130473158	0.25880310289133088	4414.04	2.84	108.91311874410560	0.0452048828242952	0	
ES091112449565	246	Mon Nov 22 22:52:00 CET 2021	16.4	671.0	134.73	3133.5	0.11		6	676.45424245194419	24473.107192732232	0.5486701894738021	4109.94	2.15	146.40323741909095	0.091117815297770	0	

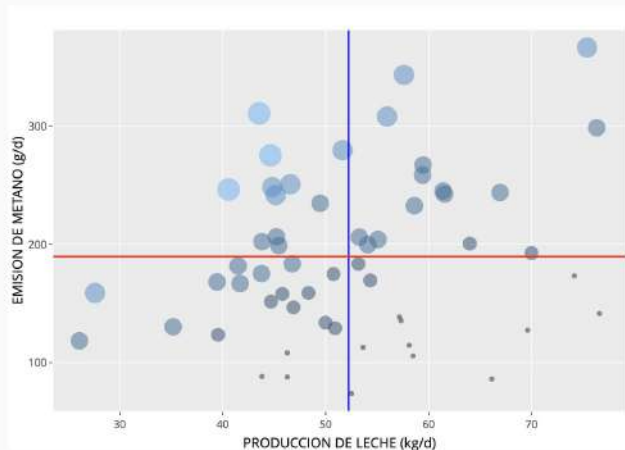
es	numero	collar	nupadre	numpar	DEL	metodo	leche	grasa	proteina	lactosa	fcontrol	fsar	cib	ultimaCub	DespuesUI	Parto	PesoVivo
12	ESPH15044697451	4663	ESPM1504153910	1	318	n1	49.90	3.62	3.26	5.01	07/12/2021 0:00:00	23/01/2021 0:00:00	ES061112154663	10/09/2021 0:00:00			890.925
12	ESPH1504739508	9490	840M3130915852	1	170	n1	49.70	2.19	3.21	5.01	07/12/2021 0:00:00	20/06/2021 0:00:00	ES051112449480	24/08/2021 0:00:00			657.413
12	ESPH1504739622	9478	840M3130915852	1	98	n1	40.50	3.23	3.16	5.21	07/12/2021 0:00:00	31/08/2021 0:00:00	ES041112449478	04/11/2021 0:00:00			578.523
12	ESPH1504739623	9479	840M3130915852	1	274	n1	35.50	3.32	3.47	4.85	07/12/2021 0:00:00	08/03/2021 0:00:00	ES051112449479	21/05/2021 0:00:00			647.666
12	ESPH1504745568	4892	CANM0011993494	1	482	n1	19.20	4.44	4.47	4.73	07/12/2021 0:00:00	12/08/2020 0:00:00	ES021112194692	24/06/2021 0:00:00			590.588
12	ESPH15047481420	4723	ESPM9204487906	1	339	n1	28.70	4.53	3.97	4.66	07/12/2021 0:00:00	02/01/2021 0:00:00	ES081112194723	21/05/2021 0:00:00			615.612
12	ESPH1504787248	9492	GBFM6386050583	1	339	n1	33.90	4.55	3.79	5.02	07/12/2021 0:00:00	02/01/2021 0:00:00	ES061112449492	21/05/2021 0:00:00			658.032
12	ESPH1504790998	9499	840M3008461593	1	382	n1	34.90	3.22	3.40	4.99	07/12/2021 0:00:00	20/11/2020 0:00:00	ES031112449499	18/11/2021 0:00:00			623.461
12	ESPH1504790999	9501	840M3008461593	1	291	n1	38.90	4.10	3.51	4.70	07/12/2021 0:00:00	19/02/2021 0:00:00	ES021112449501	08/07/2021 0:00:00			632.368
12	ESPH1504795200	9514	840M3014562212	1	223	n1	33.60	4.42	3.79	4.99	07/12/2021 0:00:00	28/04/2021 0:00:00	ES041112449514	10/09/2021 0:00:00			641.056
12	ESPH1504828783	9522	840M3130915852	1	309	n1	40.50	2.22	3.34	4.97	07/12/2021 0:00:00	01/02/2021 0:00:00	ES011112449522	21/04/2021 0:00:00			657.193
12	ESPH1504828785	9524	840M3132349851	1	257	n1	30.20	4.73	3.69	5.10	07/12/2021 0:00:00	25/03/2021 0:00:00	ES031112449524	30/05/2021 0:00:00			564.904
12	ESPH1504828786	9525	840M3130915852	1	305	n1	38.00	3.44	3.38	5.04	07/12/2021 0:00:00	05/02/2021 0:00:00	ES041112449525	20/04/2021 0:00:00			673.949
12	ESPH1504828787	9529	840M3132349851	1	270	n1	40.70	4.46	3.79	4.99	07/12/2021 0:00:00	12/03/2021 0:00:00	ES081112449529	28/05/2021 0:00:00			632.888
12	ESPH1504828788	9531	840M3132349851	1	1403	n1	48.90	3.00	3.76	5.14	07/12/2021 0:00:00	27/09/2021 0:00:00	ES061112449531	25/11/2021 0:00:00			621.628

We provide a report to each farm

# Report

## // 1.2. Emisión de metano expresada en gramos/día (g/d) por producción de leche.

En la siguiente grafica se muestra la producción de metano en función (gramos/día) de la producción de leche (kg/d) de las vacas de la ganadería dividida en 4 cuadrantes, siendo aquellas que menos metano emiten y producen mayor cantidad de leche las situadas en el **cuadrante inferior derecho**.



Se divide la tabla en 4 cuadrantes, la línea horizontal (roja) indica la media de la ganadería para emisiones de CH<sub>4</sub> y la línea vertical (azul) indica la media de la producción de leche de la ganadería.

## // 2. DIFERENCIA ENTRE LAS VACAS CON MENORES Y MAYORES PERDIDAS POR METANO POR LITRO DE LECHE PRODUCIDO:



La diferencia entre el 10% de vacas con mayores y menores pérdidas por emisiones fue de **3.86 g de METANO por cada kg de leche producido**

POSIBILIDADES DE MEJORA DEL APROVECHAMIENTO DE LA RACION: Esta diferencia equivale a 0.028 UFL pérdidas por litro de leche que es la energía que proporcionan 28 gramos de pienso, lo que supondría un total de 538 kg de pienso por vaca al año.



### 3. RESULTADOS DE LAS VACAS DE LA GANADERIA EN FUNCION DE LA EMISION DE METANO POR KG DE LECHE (GRAMOS AL DIA POR LITRO DE LECHE).

La siguiente tabla muestra el ranking de las vacas de la ganaderia **ordenadas segun menores emisiones de METANO por cada kg de leche** producido (el valor por cada vaca está corregido por días en lactacion, numero de lactacion, mes de parto y robot).

numero	n.establo	n.parto	dias en leche	EMISIONES DE METANO (g/d)	EMISIONES DE METANO POR KG DE LECHE	RANKING POR EXPLOTACION
ESPH1703944464	251	3	152	100.88	1.69	10% MENOS EMISIONES POR KG DE LECHE
ESPH1704129715	293	1	383	139.48	1.75	10% MENOS EMISIONES POR KG DE LECHE
ESPH1704365573	317	1	232	131.59	1.88	10% MENOS EMISIONES POR KG DE LECHE
ESPH1703611386	200	4	173	101.79	2.15	10% MENOS EMISIONES POR KG DE LECHE
ESPH1704129714	292	2	145	235.82	5.29	10% MAS EMISIONES POR KG DE LECHE
ESPH1704371619	336	1	56	313.08	5.49	10% MAS EMISIONES POR KG DE LECHE
ESPH1704129712	290	2	27	291.49	5.67	10% MAS EMISIONES POR KG DE LECHE
ESPH1704323433	311	2	12	186.24	6.04	10% MAS EMISIONES POR KG DE LECHE
ESPH1704236543	301	2	21	298.76	6.96	10% MAS EMISIONES POR KG DE LECHE

### 4. INDICES GENETICOS TOROS

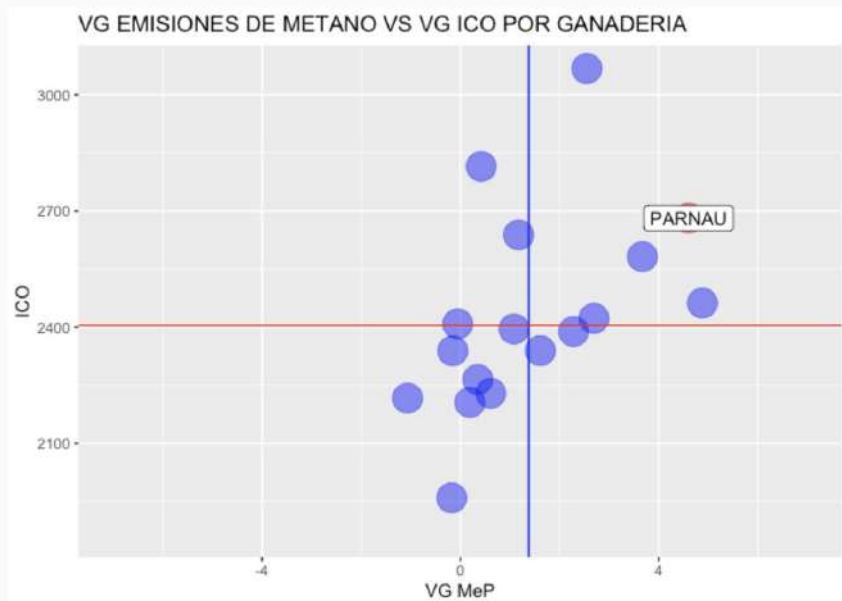
La siguiente tabla muestra la valoracion genetica de los toros con hijos/as en la ganaderia para los indices ICO, ICAP, y kg de leche, grasa y proteina, y emisiones de metano expresadas en produccion, g/d (MeP) y en concentracion, ppm (MeC).

nombre	animal	ICO	MeC	MeP	kl	kg	kp	ICU	ICAP
COOKIECUTTER MOM HUNTER ET	840M3000540481	2958	31.22	7.80	663	34	33	0.63	0.51
LONG-LANGS OMAN OMAN 2 ETN	840M3006030062	2764	107.63	17.13	332	39	40	0.17	-0.16
S-S-I MONTROSS JEDI ET	840M3123886035	4237	-2.94	-0.13	1942	32	72	2.08	0.70
GILLETTE WINDBROOK ET	CANM0007816429	2420	-0.72	-1.67	158	33	16	1.00	1.68
MAPLE-DOWNS-I G W ATWOOD ET	CANM0008956379	2762	53.83	8.22	-22	28	6	2.07	1.16
GENERATIONS EPIC ET	CANM0011104016	3395	103.25	21.09	810	27	31	1.42	0.83
SNOWBIZ LITTLETON ETM	CANM0108251028	3475	14.78	7.90	2031	50	51	1.03	0.26
BURANO ET	DEUM0357970606	3380	36.14	5.92	917	2	32	2.05	0.32

# Report

## // 5.2. EMISIONES DE METANO - ICO

En la siguiente grafica se muestra la comparativa entre explotaciones para la media del valor genetico de ICO y la media del valor genetico de emisiones de metano. Siendo el **objetivo** situarse en el **CUADRANTE SUPERIOR IZQUIERDO**.



# Data processing

Java Program	Function
SnifferAnalyzer.jar	1. Assign events to cows according to time footprint and lag
	2. Calculate background (average of 5 lowest measurements from opening of the AMS gate to cow exit).
	3. Detect eructation peaks
	4. Calculate traits
	5. Write output

# Data processing

R script	Function
	Combine output from SnifferAnalyzer with test day records and ID information