How genetic selection could contribute to the multi performance of farming systems?

A modelling approach applied to Lacaune dairy sheep

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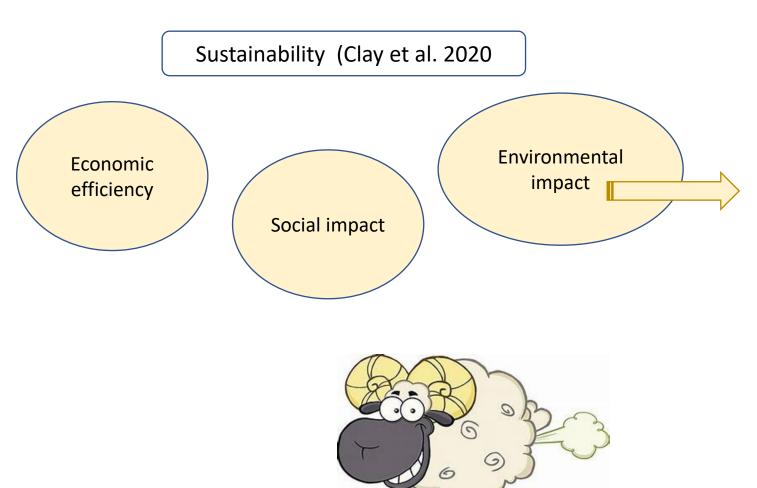


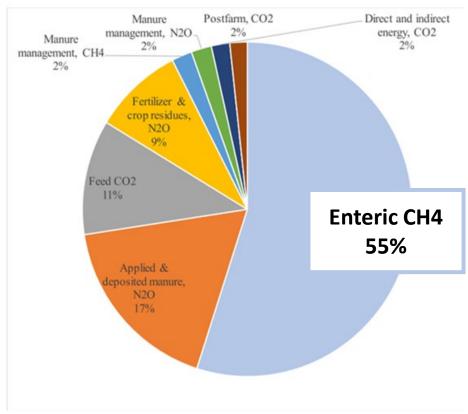






## Multiperformance of farming systems





Global GHG emissions from small ruminant supply chains; adapted from Gerber et al. (2013).



### Improvement approaches

#### Multiperformance

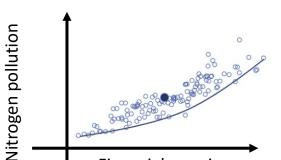
# Management practices Farming system design

Feed, reproduction

#### **Selection practices**

- Selecting traits that enhance both productivity and sustainability
  - Operates at population levels
  - Focusing on long-term changes

→ use of **individual-based models** to manage trade-offs at farm level



Ex: in dairy sheep Villalba et al. 2019

Lacaune breed

Current breeding goal

Milk yield

Milk composition

**Udder score** 

Somatic cell score

→ no selection ongoing for environmental performances (e.g. CH4)



Financial margin

### Approach of this study

Individual-based modelling to predict responses to variation in production environments and to selection

→ trade-offs at the individual level based on resource allocation theory

Examples:

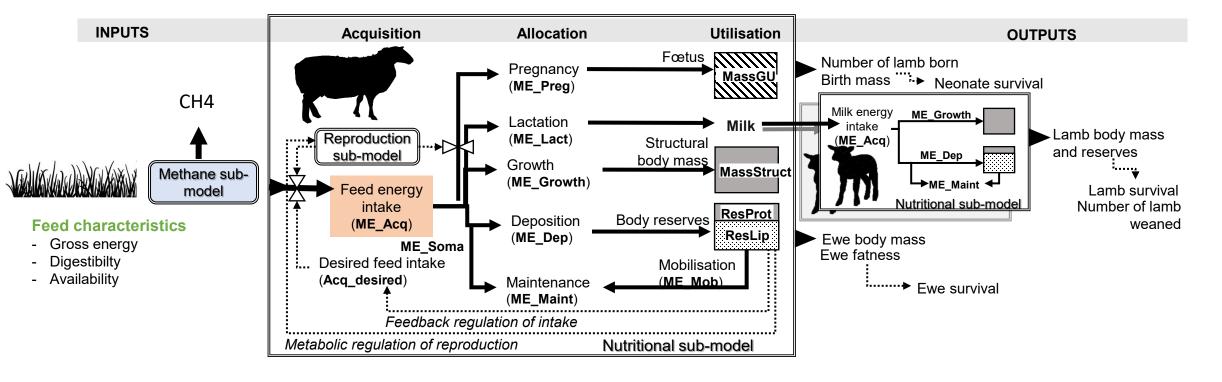
Dairy goats: Douhard et al. 2014

Dairy cows: Puillet et al. 2016, 2021; Bouquet et al. 2024

**Aim**: Simulate phenotypic variability of the output traits using an individual-based model of dairy sheep



# Sheep model

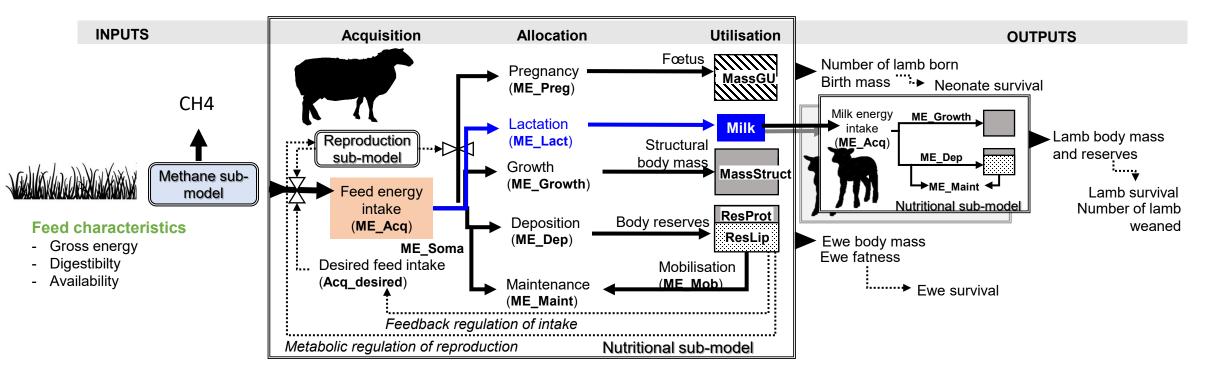


Metabolizable energy intake (ME\_Acq)

= Feed gross energy - Gaseous energy (CH4) - Urinary energy - Fecal energy



### Focus 1: simulation of individual variability

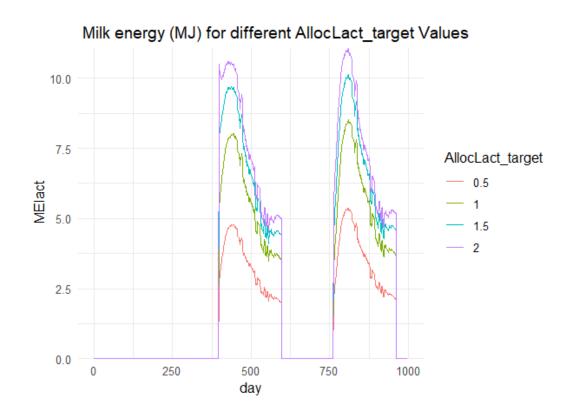


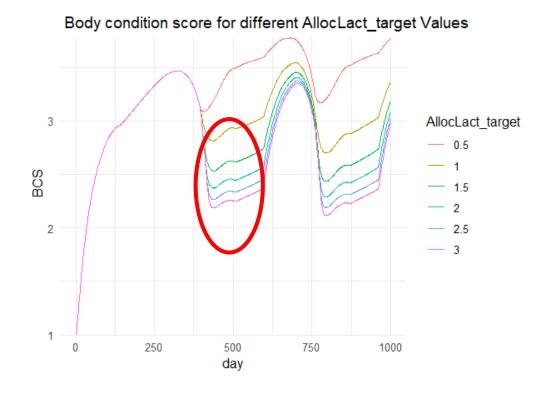




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### Focus 1: simulation of individual variability

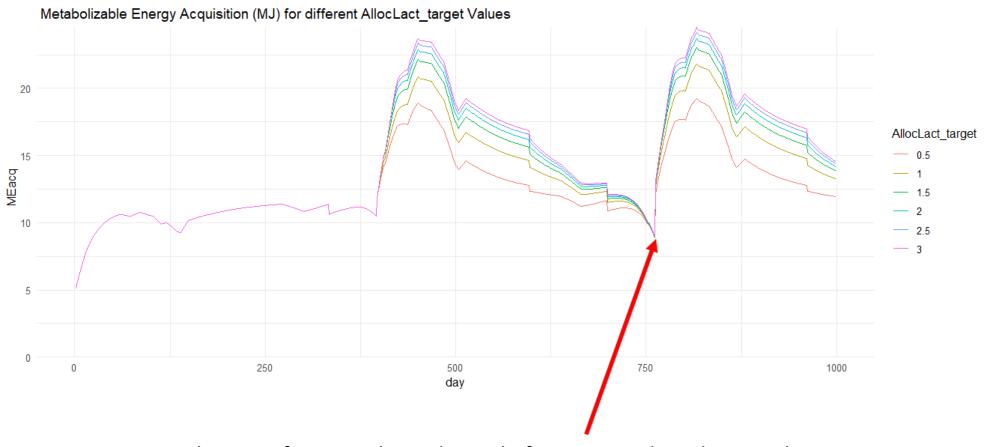




→ AllocLact<sub>target</sub> can mediate a trade off among individuals between milk and BCS



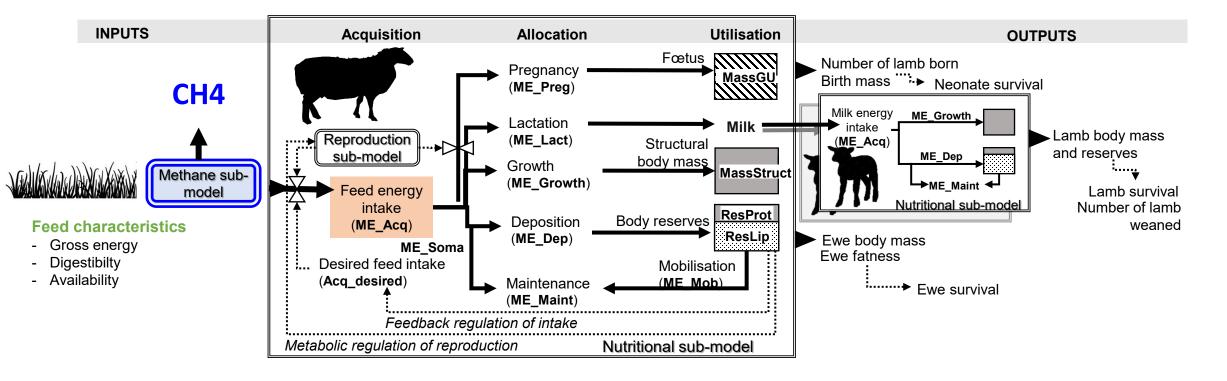
## Focus 1: simulation of individual variability



Reduction of ewe intake at the end of gestation when the gravid uterus mass is high (Wilkinson and Chestnutt 1988)



### Focus 2: predict enteric CH4 emissions



Metabolizable energy intake (ME\_Acq)

= Feed gross energy - Gaseous energy (CH4) - Urinary energy - Fecal energy



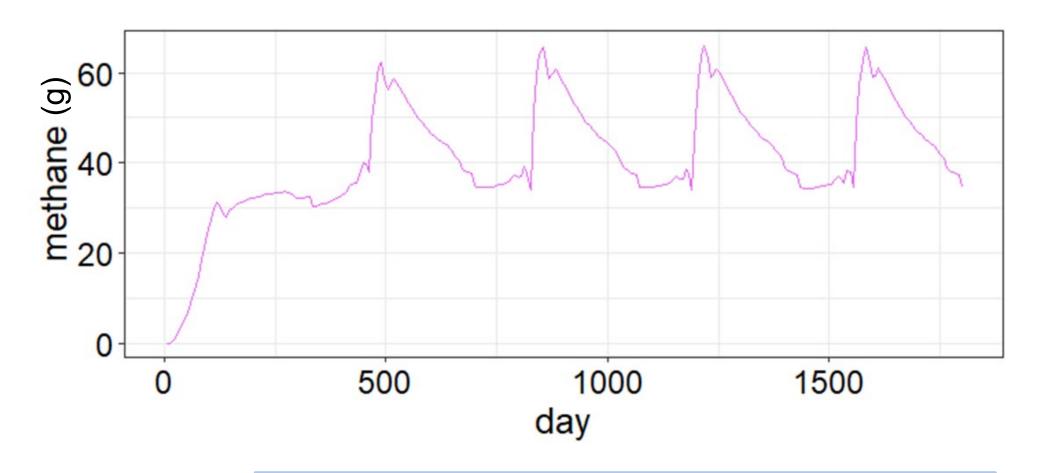
#### Focus 2: Predict enteric CH4 emissions

CH4 = DOM × (45.42 – 6.66 · 
$$\left(\frac{\text{Feed Intake}}{\text{Bodyweight}}\right)$$
 + 0.75  $\left(\frac{\text{Feed Intake}}{\text{Bodyweight}}\right)^2$  + 19.65 · PCO – 35.0 · PCO<sup>2</sup> – 2.69 ·  $\left(\frac{\text{Feed Intake}}{\text{Bodyweight}}\right)$  · PCO)

- Digestible organic matter
- Proportion of concentrate in feed

Sauvant et al. 2011

#### Focus 2: Predict enteric CH4 emissions



→ Within-individual changes in enteric CH4 emissions can be predicted



### Conclusions & the next steps

#### Multiperformance

#### (1) Individual variability

- → Estimate variance of coefficients using real data in Lacaune dairy sheep
- → Include a genetic component

Example: h<sup>2</sup> AllocLact<sub>target</sub>

Bouquet et al. 2024

#### (2) Prediction of CH4 emissions

- → Simulate individual variability in CH4 and potential trade-offs with production traits
- → Carbon footprint at the farm level

→ An approach to explore how genetic selection could contribute to the multi performance of farming systems







