



Zootechnie Prediction of the daily nutrient requirements of gestating sows based on behavioural sensor data and machine-learning algorithms

















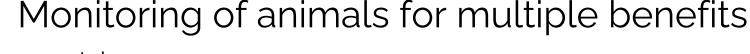
M. Durand¹, C. Largouët², L. Bonneau², J.Y. Dourmad¹, C. Gaillard¹

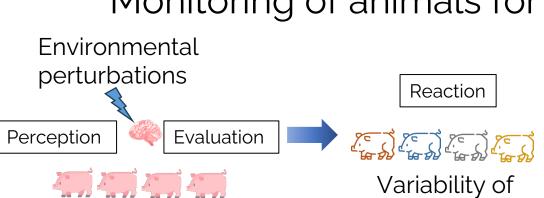
¹PEGASE, INRAE, Institut Agro, 35590, Saint Gilles, France ²Institut Agro, Univ Rennes1, CNRS, INRIA, IRISA, 35000, Rennes, France



Context

Monitoring of animals for multiple benefits





Variability of responses



Measured





Sensors or connected tools

Large amount of data



Durand M., Largouët C., Bonneau de Beaufort L., Dourmad J.Y., Gaillard C., 2023. A dataset to study group-housed sows' individual behaviours and production responses to different short-term events. Animal Open Space, 2, 100039

Identify changes in welfare status



Durand M., Largouët C., Bonneau de Beaufort L., Dourmad J.Y., Gaillard C., 2023. Prediction of gestating sows' welfare status based on machine learning methods and behavioural data. Sci. Rep., 13, 21042

Identify changes in behaviour



Durand M., Dourmad J.Y., Julienne A., Couasnon M., Gaillard C., 2023. Effects of a competitive feeding situation on the behaviour and energy requirements of gestating sows. Appl. Anim. Behav. Sci., 261, 105884



Identify changes in nutrients requirements

Crédit Photo: Copilot, Durand et al., 2023 2

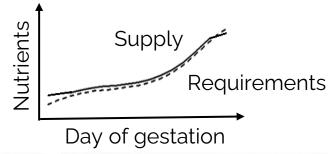
Durand M., Largouët C., Bonneau de Beaufort L., Dourmad J.Y., Gaillard C., 2023. Prediction of daily nutritional requirements of gestating sows based on sensors data and machine learning algorithms. J. Anim. Sci., 101, skad337



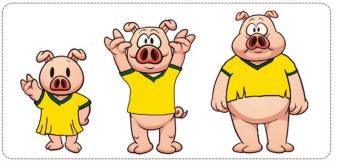
Context

Precision feeding for gestating sows

Precision feeding aims to bring the proper amount of feed with the suitable composition to be supplied in a timely manner to individual animals (Pomar et al., 2009).













Reduction of underfed and overfed animals

Decrease of nitrogen excretion by 18 % ans phosphorous by 9 %

Decrease of lysine intake by 25 % and phosphorous by 8 %



Aim of the study



Mechanistic model designed for precision feeding of gestating sows: daily metabolizable energy and standardized ileal digestible lysine requirements for each sow

Difficulty of using the model to estimate requirements



Search for a new method made possible by the development of sensors

Explore and evaluate the prediction of individual daily nutrient requirements, using data measured by sensors and machine learning algorithms



73 gestating sows (4 batches)





9 machine-learning algorithms



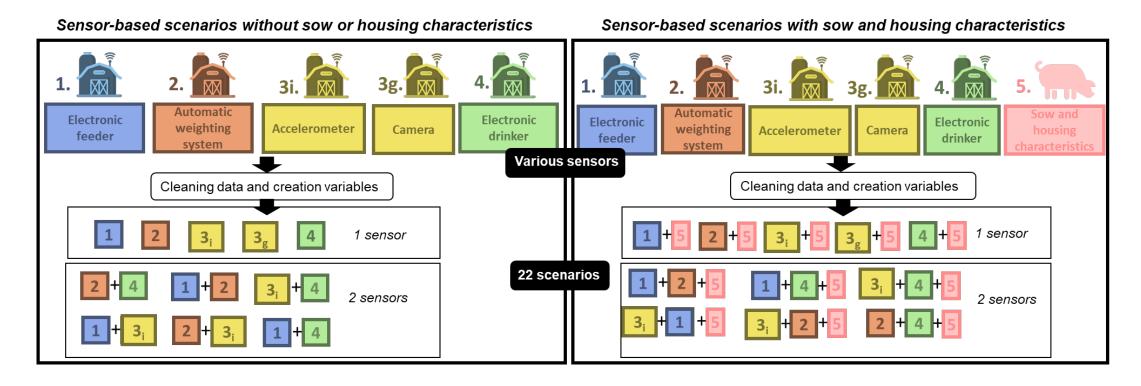
Various digital farm configurations (electronic feeder, drinker stations)



Material and methods (1/2)



Step 1: Scenarios definition



Material and methods (2/2)



Step 1: Scenarios definition





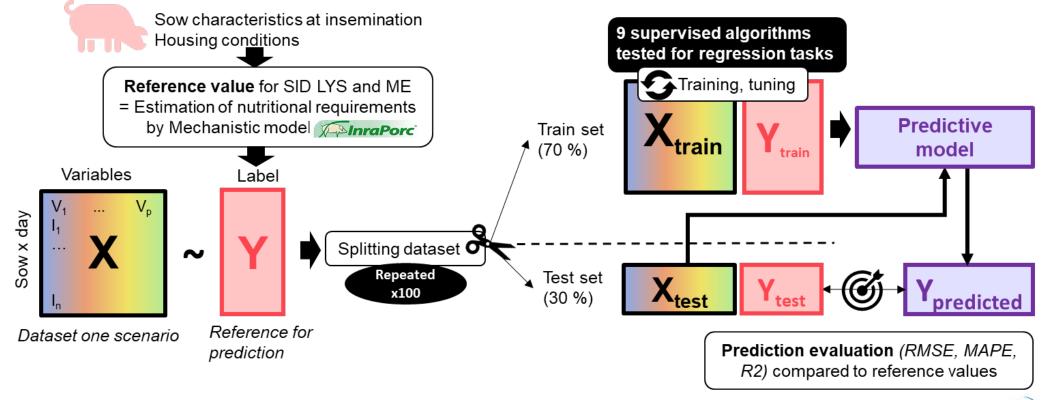




10 Scénarios
Sensors + Sow & housing characteristics (SP.)

10 Scénarios Sensors data only (S.)

Step 2 : Prediction of nutrients requirements with 9 machine learning algorithms



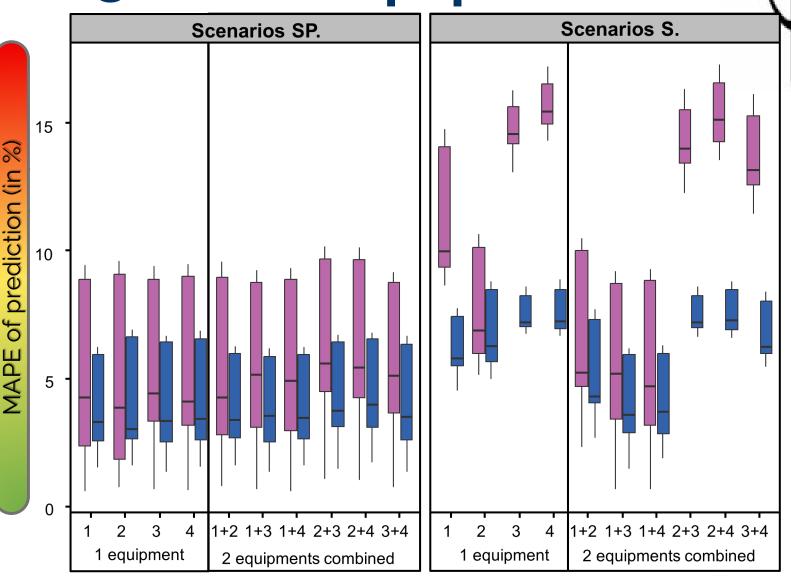
Main results (1/3): best equipments

Metabolisable energy

Best performance obtained for scenarios with **electronic feeder alone** (R² = 0.74) **or combined** (R² = 0.75)

Best performance obtained for scenarios with **electronic feeder or automatic weighting system** (R² = 0.92)

- 1: electronic feeder
- 2: automatic weighting system
- 3: accelerometer, camera
- 4: electronic drinker



Digital farm configurations



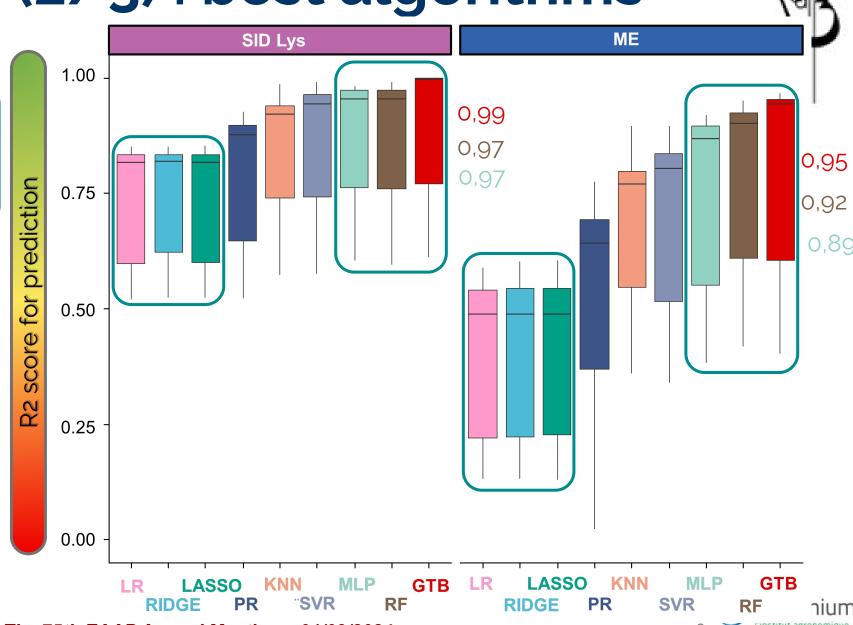
Main results (2/3): best algorithms

Best performance for GTB

(MAPE = 0.91 MJ/d for energy and 0.08 g/d for lysine), followed by **RF and MLP**

Worst performance for linear regressions (MAPE = 2.75 MJ/d for energie and 1.07 g/j pour lysine)

- Linear Regression (LR)
- LR with a RIDGE regularization (RIDGE)
- LR with a LASSO regularization (LASSO)
- Polynomial Regression (PR)
- k-nearest-Neighbors (KNN)
- Support Vector Machine for Regression (SVR)
- Multilayer Perceptron (MLP)
- Random Forest (RF)
- Gradient Tree Boosting (GTB)



Main results (3/3): ablation study

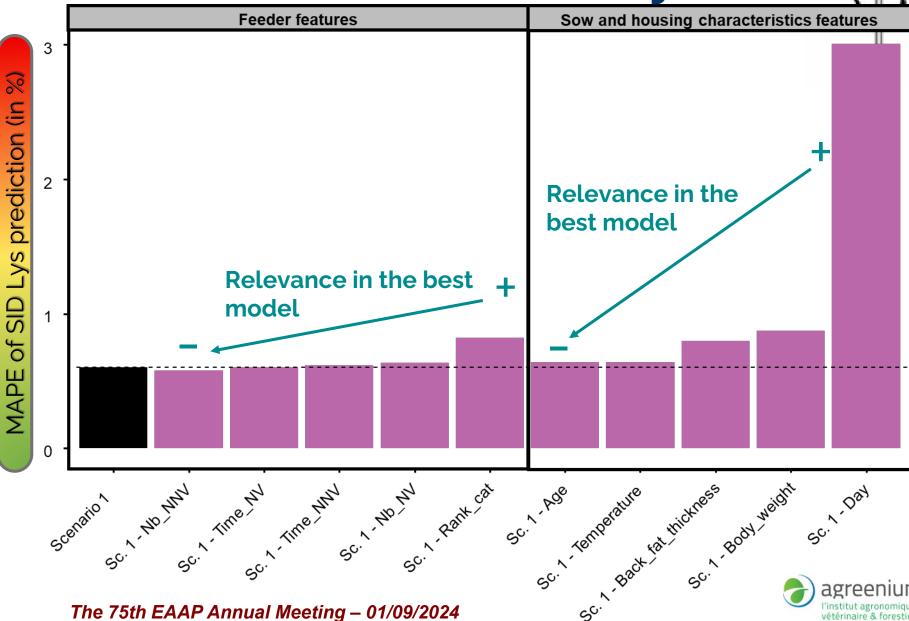
Inclusion of sow and housing characteristics

reduced the MAPE by 2.2 pts for energy and 5.6 pts for lysine

Relevance of features:

Importance of feeder order, number of nutritive visit, day of gestation, BW and back-fat-thickness

NV : nutritive visit to feeder NNV : non-nutritive visit to feeder



Conclusion

Prediction of the daily nutrient requirements of gestating sows based on behavioural sensor data and machine-learning algorithms



- Very good prediction performance with low errors (less than 5 and 7% for energy and lysine)
- Superiority of the electronic feeder among the tested scenarios
- Importance of including sow and housing characteristics to improve prediction
- Superiority of decision tree machine learning methods

Thank you for your attention