Data-Driven System for Vertical Farming and Milk Quality Prediction: A Framework for Investigating the Impact of Microgreens in Dairy Nutrition

ANDRAS REVOLY¹, ISTVÁN SZABÓ¹, JÁNOS TŐZSÉR², BENCE TARR¹

¹Institute of Technology, Hungarian University of Agriculture and Life Sciences ²Faculty of Agricultural and Food Sciences, University of Győr

Introduction

Precision milk-component pricing and the growing demand for premium-quality dairy products place renewed focus on smart feed design.

To quantify this potential before launching costly in-vivo trials, we assembled two, independently validated building blocks:

Machine-learning (ML) estimator – Trained on 25 000 commercial milk-record pairs, the multi-output model predicts casein concentration and somatic cell count (SCC) from routinely monitored milk traits (fat, protein, lactose, urea, conductivity, etc.).

Sensor-rich growth chamber – A custom vertical-farm module equipped with PAR, CO₂, temperature, humidity and EC sensors, delivering microgreen batches with repeatable nutrient profiles at <4 kWh kg⁻¹ DM.

Our working hypothesis is that these two assets can be wired into a closed loop: fixed-composition microgreens are blended into the basal ration at graded inclusion levels, the ML estimator forecasts the resulting casein yield and SCC trajectory.

Materials and Methods

Crop-production platform

A purpose-built, cloud-connected vertical-farm chamber supplies the microgreens.

Lighting – Programmable ePAR LED array (400–750 nm) with six independently dimmable channels enables bespoke light recipes and light programs. Power draw is continuously modulated to maintain a target PPFD of 250–300 μ mol m⁻² s⁻¹ at canopy height.



BIOSYSFOODENG 2025





Climate & hydroponics – Recirculating deep-water culture with automatic pH and EC regulation keeps the nutrient solution stable; a thermostated air loop plus mist cooling clamp the vapour-pressure deficit at 0.9 kPa.



The measured microgreen nutrient profile is merged with a library of conventional forages (maize silage, alfalfa haylage) and concentrates (soybean meal, ground maize). Microgreens replace these ingredients in 5 percentage-point steps from 0 % to 40 % of dietary dry matter while the remaining ration is re-balanced to stay isoenergetic.

SSC and casein estimation from easily measurable milk constituents:

- Baseline: multiple linear regression (scikit-learn).
- Advanced: feed-forward neural network (TensorFlow/Keras) with two hidden layers (64-64 neurons, ReLU activations) and an Adam optimiser (learning-rate 0.001, 100 epochs, batch 32).

Both models were fitted to 25 000 monthly milk-feed records collected over three years from three Hungarian Holstein herds (80 % train, 10 % validation, 10 % test). Target variables were total protein, casein, log-transformed SCC and milk oleic-acid concentration.

Results

Cross-validation yielded $R^2 = 0.86$ and RMSE = 0.12 g 100 g⁻¹ for casein; $R^2 = 0.79$ and RMSE = 0.18 for log-SCC. The neural-network marginally out-performed linear regression ($\triangle RMSE \approx 4\%$), especially at high microgreen inclusion (> 25 % DM).

Trait	Linear regression	Neural network	∆RMSE
Milk protein (%)	R ² = 0.86, RMSE = 0.12	R ² = 0.92, RMSE = 0.08	-33 %
Casein (%)	R ² = 0.84, RMSE = 0.11	R ² = 0.91, RMSE = 0.07	-36 %
Log-SCC	$R^2 = 0.64$, RMSE = 0.24	R ² = 0.79, RMSE = 0.18	-25 %
Oleic acid (g 100 g ⁻¹)	R ² = 0.75, RMSE = 0.015	R ² = 0.88, RMSE = 0.010	-33 %



(((• 00000



The feed-forward neural network consistently out-performed multiple linear regression, capturing non-linear diet-milk relationships across all four target variables.

Results of the simulated microgreen inclusion (0–40 % of dietary DM):

Milk protein: 3.25 → 3.60 % (+11 %) Casein: 2.55 → 2.95 % (+16 %) Log-SCC: 5.25 \rightarrow 5.05 (\approx -20 000 cells mL-¹) Oleic acid: 0.28 \rightarrow 0.32 g 100 g⁻¹ (+14 %)

The above results came from the most profitable ration (+30 % microgreens).





Conclusion

We now possess every functional element required for a closed-loop, precision-nutrition system: (i) a low-energy vertical-farm chamber that reproducibly delivers microgreens of known nutrient density, and (ii) a rigorously validated machine-learning model that converts feed profiles into quantitative milk-quality predictions.



