

# Exploring machine learning algorithms on activity and feeding behaviour for early estrus detection in dairy cows



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## Background

The economics of milk production and herd replacement are directly affected by the level of reproductive performance.

The recent advances in sensor technology have allowed accurate predictions of estrus events using animal behavior information.

Individual cow sensors (MooMonitor+, DairyMaster, Co. Kerry, Ireland) can monitor cow neck movements 24/7 for heat related activity, rumination, resting, feeding, head position and restlessness. This system improves farm profitability by decreasing labour requirements for farm personnel, improving reproductive performance and minimising losses due to missed heats, undiagnosed illnesses and general animal cow health.



**MooMonitor+ (DairyMaster, Co. Kerry, Ireland)**

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## Data and Analysis

The objective of this study was to evaluate the potential of machine learning techniques to identify the onset of estrus in dairy cows using hourly behaviour data collated from 1,750 dairy cows housed in a free-stall barn at 5 commercial dairy farms during the 2 years period. The behaviour was measured using the behaviour-monitoring collar (MooMonitor+, DairyMaster, Causeway, Co. Kerry, Ireland) of each animal 24 h/day. The MooMonitor+ recording system stores information every minute and then summarizes this data on an hourly basis.

### Original variables (hourly data):

Cow ID: int  
Date: datetime64[ns]  
DIM - int  
Heat (0,1)  
Calving date - multiple per cow  
Lactation - how many calves the cows have had int 1...n  
+Rumination - int (mins spent rumination in an hour)  
+Resting int (mins spent resting in an hour)  
+Feeding int (mins spent feeding in an hour)  
+High activity - int (mins spent rumination in an hour)  
+Med activity - int (mins spent rumination in an hour)  
+Low activity int (mins spent rumination in an hour)

### Calculated variables:

DSLH = days since last heat, (eg. num days between heats)  
mean DSLH = mean OF ALL period between heats per cow

X = df[['DSLH', 'meanDSLH', 'High\_27 norm', 'Medlow\_27 norm', 'High\_6 norm', 'Medlow\_6 norm', 'High\_9 norm', 'Medlow\_9 norm', 'High\_12 norm', 'Medlow\_12 norm', 'High\_15 norm', 'Medlow\_15 norm', 'High\_18 norm', 'Medlow\_18 norm', 'Rum\_21 norm', 'High\_21 norm', 'Medlow\_21 norm', 'Rest\_24 norm', 'High\_24 norm', 'Medlow\_24 norm', 'High\_72 norm', 'Medlow\_72 norm', 'High\_120 norm', 'Medlow\_120 norm', 'High\_168 norm', 'Medlow\_168 norm', 'High\_240 norm', 'Medlow\_240 norm', 'High\_360 norm', 'Medlow\_360 norm', 'High\_432 norm', 'Medlow\_432 norm']]

Different classification machine learning techniques (Extreme Gradient Boosting (GBDT), Random Forest (RF), Decision Tree (CART), K-Nearest Neighbors (KNN), Gaussian Naïve Bayes (NB), Linear Discriminant Analysis (LDA), Logistic Regression (LR)) were evaluated to identify the onset of estrus in dairy cows.

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## Results

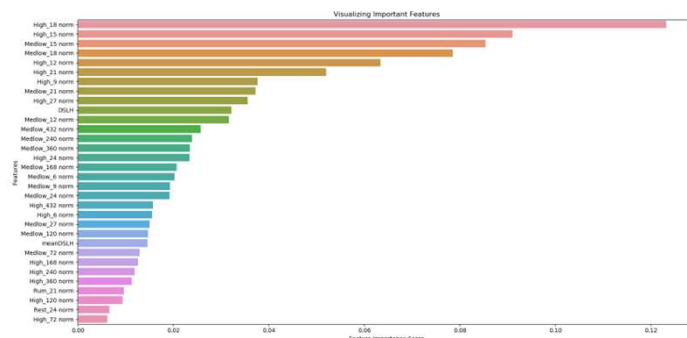
- The study showed that the accuracy, precision, sensitivity, specificity and F1 score of the Random Forest algorithm were up to 94.98%, 99.19%, 94.62%, 96.60%, 96.85%, respectively, which indicates a great potential for early estrus detection.

		Predicted Class		
		Positive	Negative	
Actual Class	Positive	183662	1432	Sensitivity $\frac{TP}{(TP + FN)}$
	Negative	10363	42112	Specificity $\frac{TN}{(TN + FP)}$
		Precision $\frac{TP}{(TP + FP)}$	Negative Predictive Value $\frac{TN}{(TN + FN)}$	Accuracy $\frac{TP + TN}{(TP + TN + FP + FN)}$

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## Conclusion

- To enhance reproductive performance in dairy cows, it is essential to detect estrus onset associated with ovulation time in a timely and accurate way to ensure high pregnancy rates and profitability.
- Behavioral data about dairy cows' rumination, resting, feeding, and activity were utilized in our model. These data were collected by DairyMaster's MooMonitor+, which is the first system validated in both indoor and outdoor systems. The model's accuracy was improved by including calculated variables like days since last heat and mean of days since last heat. Different rolling means were calculated after normalizing behavioral data to the herd level. Finally, the most important features for the model were selected and parameters were changed in the algorithm appropriately.



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