



Time series data analysis to predict the status of mastitis in dairy cows by applying machine learning models to automated milking systems data

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#### Introduction

- ➤ Mastitis is a significant disease affecting dairy cows, leading to economic losses and welfare concerns
- > Early detection and management of mastitis are crucial for:
  - > minimizing the risks to animal health & welfare > maximizing economic gains
- Traditional detection methods rely on manual inspections, which are laborintensive
- ➤ Automated Milking Systems (AMS) generate large datasets that can be leveraged using Machine Learning (ML) for early mastitis prediction
- The aims of this study were to apply ML techniques to AMS data to predict mastitis occurrence:
  - > one day prior to its observation and
  - > on the day of mastitis observation.

### Materials & methods

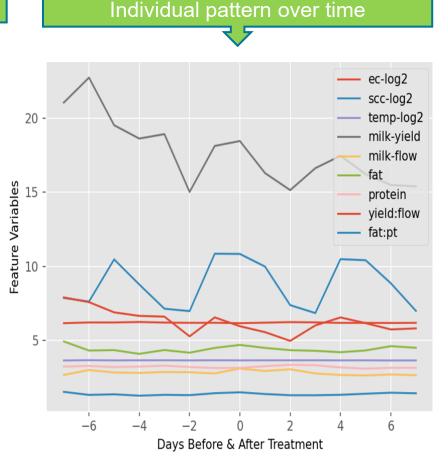


- Original data: AMS milk records 3-4 times per day per cow: 2.73 million observations
- Number of Individual Cows: 1790
- Number of Farms: 2
- Time Duration: 2019-2022 (4 years)
- X variables 7 (details on next slide...)
- Y variable 1
- Time resolution adjusted to one time (average value per day per cow)
- Total Observations 953,270
- Total negative obs. 942290 (days without treatment)
- Total Positive obs. 2250 (days with treatment)

# Data format tranformation (auto-regressive order)

#### Seven predictor (X) variables

- 1. E-conductivity
- 2. Somatic cell count
- 3. Milk temperature
- 4. Milk yield
- 5. Milk flow
- 6. Fat content
- 7. Protein content



#### Auto-regressive data format econd- econdconduct 70.33 72.75 72.00 73.00 74.25 69.67 70.33 72.75 72.00 73.00 69.67 70.33 71.00 72.75 72.00 71.00 69.67 70.33 72.75 73.67 70.33 73.67 71.00 69.67 70.33 xt + xt-1 + xt-2 + xt-3 + xt-4

## **Model Formats**

Format-1: Mastitis prediction one day prior

Predictors (from AMS)

- Somatic cell count
- E-conductivity
- · Milk temperature
- Milk yield
- Milk flow
- Fat content
- Protein content

Date of status change = recorded treatment

Individual pattern over time (xt-1 + xt-2 + xt-3 + xt-4)

Each row contains values of all predictor variables 4 days before the day of current status

Format-2: Mastitis prediction on same day of observation



- Somatic cell count
- E-conductivity
- Milk temperature
- Milk yield
- Milk flow
- Fat content
- Protein content

Date of status change = recorded treatment

Individual pattern over time (xt + xt-1 + xt-2 + xt-3)

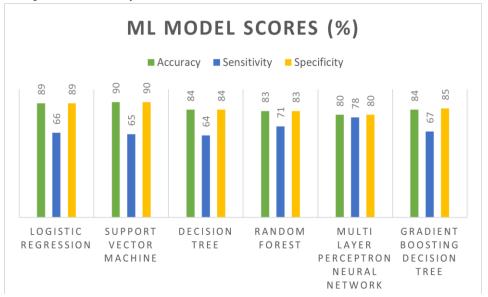
Each row contains values of all predictor variables (current + 3 days before) the day of current status

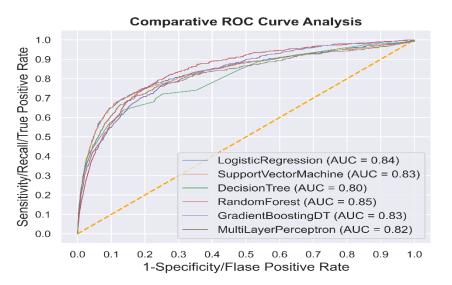
## Training and testing of ML models

- ➤ Dataset was split into: >training (75%) and >testing (25%)
- > Training dataset (highly class imbalance) > Negative = 722662 > Positive = 1800
- Synthetic minority oversampling technique (SMOTE)
  - ➤ After oversampling >Negative = 722662 >Positive = 722662
- > Testing dataset (No oversampling) > Negative = 208299 > Positive = 450
- ➤ Six ML models applied:
  - > Logistic regression (LR) > Support vector machines (SVM) > Decision tree (DT)
  - > Random forest (RF) > Gradient boosting decision tree (GBDT)
  - > Multi-layer perceptron neural network (MLP-NN)
- > GridSearchCV: > Hyper-parameter tuning > 5-fold cross validation

## Results part-1 (mastitis prediction one day before)

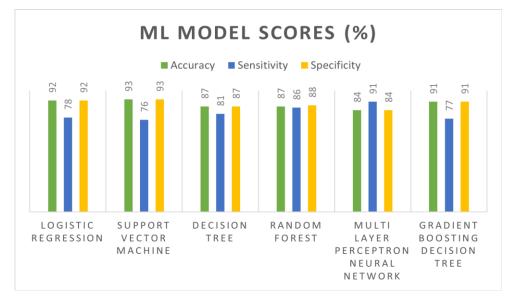
- Over all
  - ➤ Accuracy 80-90%
  - > Sensitivity 64-78%
  - Specificity 80-90%
- Highest accuracy and specificity > support vector machine
- Highest sensitivity > multi layer perceptron
- Overall moderate predictive accuracy
- > Area under curve (AUC) score 0.8-0.85
- Random forest classifier had greatest AUC
- Decision tree scored lowest

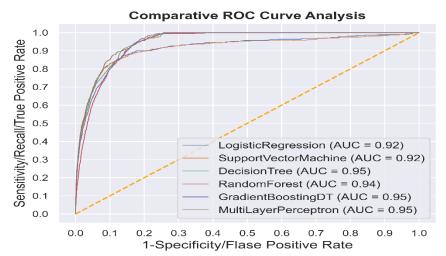




## Results part-2 (mastitis prediction on the day of observation)

- Over all
  - ➤ Accuracy 84-93%
  - ➤ Sensitivity 76-91%
  - ➤ Specificity 84-93%
- Highest accuracy and specificity > support vector machine
- Highest sensitivity > multi layer perceptron
- Overall all models scored an area under curve (AUC) greater than 0.9 (high accuracy)
- Decision tree, gradient boosting and multi layer perceptron had greatest AUC
- Logistic regression and support vector machine scored lowest AUC





### Discussion

- > This study developed a framework of ML models for mastitis prediction on a time series dataset generated by AMS
- Since this is an ongoing area of research, a targeted level of sensitivity and specificity for prediction models on AMS data is not yet decisively defined.
- ➤ Some studies have suggested that Mastitis detection models should have Sensitivity of ≥80% (Hillerton, 2000; Hogeveen et al., 2010, 2021)
- ➤ International Standard Organization (ISO, 2007) recommends that Sensitivity be >70% with an Specificity level of 99%.
- ➤ However, the requirements mentioned above have rarely been met to date (Khatun et al., 2018; Anglart et al., 2021; Hogeveen et al., 2021).

#### Conclusion

- Our findings indicated moderate to high accuracy of ML models to predict mastitis
- Study also demonstrated the robustness of time series AMS data to predict mastitis events in future
- ➤ However, as each model had its own strengths and weaknesses, therefore these findings have certain limitations.
- ➤ We propose inclusion of additional variables from AMS records and integration of other sensorial data for further improvement of ML models in future studies.

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