

Faculty of Agricultural and Nutritional Science

CAU

Christian-Albrechts-University Kiel

Institute of Animal Breeding and Husbandry

Lameness Detection in Sows using Accelerometer Data

C. Scheel¹, I. Traulsen¹, W. Auer², K. Müller³ J. Krieter¹

¹ Institute of Animal Breeding and Husbandry, Christian-Albrechts-Universität, Kiel ² MKW Electronics GmbH, Weibern, Austria ³ Lehr- und Versuchszentrum Futterkamp 65th Annual EAAP Meeting Copenhagen, Denmark, August 25th to 29th, 2014 Session 05, abstract number 19004, cscheel@tierzucht.uni-kiel.de





Background and objective

Why to detect lameness automatically?

- Group housing in gestation units is mandatory (per EU norm 2001/88/EG)
- Lameness is a common problem in group housing
- Increasing number of animals per farm, constantly monitor health status manually is time consuming

Objective

Predict beginning lameness automatically and reliably from analyzing an acceleration signal supplied by a monitoring system.



Ear tags (MKW Electronics)

- Positioning system
 - Using TOA (time of arrival)
 - Supplies 2-d coordinates
- Temperature sensor
 - Ambient temperature
 - Skin temperature
- Acceleration sensor
 - Supplies (x, y, z) vector
 - Programmable sample range

Data resolution vs. battery life







Floorplan Futterkamp Research Facility











Days





Data processing: Wavelet Decomposition





Outline of general method

Each sow leaves a signal *s* of accelerometer data as a record of her past behaviour.

- This should contain information on her usual (to be expected) behavioural patterns
- Find or define features (functions *f(s)* of the signal) that capture these
- Compare currently calculated feature values to the to be expected values from record
- Use unusual deviation from past record to make a prediction of beginning lameness



Feature calculation

Basic feature functions are:

- Variance (deviation from mean)
- p-variation (cumulated p-th power of differences between points)
- Features derived from histograms of data

Feature functions can be applied not only to the signal, but to its wavelet representation as well.

This results in a collection of features.



A simple model for prediction







Features on diagnosis day for 7 lame and 7 healthy sows

7 lame sows features **on** for diagnosis day

7 healthy sows matched for day and age







Features on day before diagnosis

7 sows to be diagnosed lame, features **on** before diagnosis day 7 healthy sows, matched for day and age





Results

Statistics of count of features on



- Lame/healthy were distinguishable as groups
- Individual classification would yield too many false positives/negatives



Summary and outlook

- Monitoring system is now progressed enough to produce acceleration and position data reliably (~10% data loss)
- Wavelet based feature calculation appears feasible, but:
 - We need to capture the distribution of acceleration better and thus work with smaller units of time
 - Augment feature representation with further autocorrelation features
- Too few lame samples: concentrate on representation of healthy sows' data and attempt to produce higher than usual reconstruction error on lame sows' data



Thank you for your attention!





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