

# Use of sensor data for breeding and genetics

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Using sensor data for animal health and welfare improvement  
along the dairy value chain: Merits and requirements

## Sensor data possible useful for dairy cattle breeding and genetics

### Automatic Milking Systems (AMS)

- Udder health indicators
- Fertility indicators
- BodyConditionScore (camera)
- Body weight
- Udder conformation
- Behaviour
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### + Associated sensors

- Activity meters
- 3D-camera BCS
- Rumination
- Body temperature
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## Sensor data possible useful for dairy cattle breeding and genetics

### Advantages

- Automatic, continuous recording
- Objective, frequent measures
- Less affected by herd management decisions
  
- More precise phenotypes?
- New traits that can supplement or replace current traits

### Challenges

- Different systems, providers, versions
- Access to data, data ownership
- Routine collection of data to central database

## Examples of use of sensor data for breeding and genetics

- **Cow activity** measure can be used to define new fertility traits
- Behaviour and milking efficiency traits can be derived from data from **automatic milking systems (AMS)**
- **AMS** provide udder health indicators





# Breeding for improved cow fertility

## Current fertility traits:

- Phenotypes: Insemination (AI) records and calving dates
- Success rate traits:
  - 56 d non-return
  - Conception rate
  - No of AI to conception
- Interval traits:
  - Interval from calving to first AI
  - Interval from 1<sup>st</sup> to last AI
  - Calving interval

## Challenging traits:

- Low heritability
- Need better and more precise phenotypes
  - New technology - new opportunities
    - automatic heat detection system

Cow activity measures widely used in herd management tools for heat detection

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## Aim

Examine whether new traits based on activity measures can be useful for genetic evaluation of cow fertility in Norwegian Red



Cow fertility included in the breeding program for **Norwegian Red** since the 1970's

## Data and edits

- Daily activity measure data
- 284 herds with Lely milking robot and activity tags
- 13,224 lactations
- 8,139 Norwegian Red cows
- Daily activity measures from 10 to 150 days in milk for cows with at least 50 records



# Trait



Interval (days) from **calving** to **first high activity** (CFHA)



87% of the cows had at least one episode of high activity recorded.



Mean CFHA 42 days (SD=28)



Reflects the cow's ability to return to estrus cyclicity and show heat after calving



# Heritability

$\sigma^2_a$	$\sigma^2_{pe}$	$\sigma^2_e$	$h^2$
32.14 (8.29)	13.35 (10.79)	596.16 (10.98)	0.05



## Other studies of interval to first high activity

- Løvendahl and Chagunda (2009)
  - 515 cows with records
  - $h^2$  from 0.12 to 0.18 (0.07)
- Ismael et al. (2015)
  - 3,533 cows with phenotypic records
  - $h^2 = 0.16$  (0.04)



## The trait CFHA

- showed significant genetic variation with a heritability of 0.05
- reflects the cow's ability to return to estrus cyclicity and show heat after calving
  - important aspect of cow fertility

## Other possible traits?

- Heat strength
    - Height and duration of high activity peak
  - Conception rate
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## Cow activity measurements

- can be used for genetic evaluations
- provide information on new traits that can supplement or replace current traits
- Records already exist for a large proportion of the population
- How can we best make use of these data?

All cows included?

Tags on all cows in the herd always?

# Behaviour and milking efficiency traits derived from data from AMS



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## Genetic analyses of novel temperament and milkability traits in Norwegian Red cattle based on data from automatic milking systems

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### Data:

77 herds with DeLaval  
 4883 Norwegian Red cows  
 1 mill daily records

Heritability ( $h^2$ ) with standard error (SE) of behaviour and milkability traits

Trait	$h^2$	SE
Bovertime (min)	0.27	0.03
Milking efficiency (kg/min)	0.22	0.03
Handling time (min)	0.05	0.01
Log-Handling time	0.07	0.02
Flowrate (kg/min)	0.48	0.04
Milking frequency (no)	0.05	0.01
Milking interval (hours)	0.02	0.001

## Behaviour and milking efficiency traits derived from data from AMS

Heritability ( $h^2$ ) with standard error (SE) of behaviour and milkability traits

Trait: Proportion of milkings (summarized over lactation) with	$h^2$	SE
Kick Offs	<b>0.13</b>	0.03
Incomplete milkings	<b>0.14</b>	0.03
Teat not found	<b>0.12</b>	0.03
Rejected Milkings	<b>0.05</b>	0.02

- Several traits that potentially could be used in routine genetic evaluation
- New traits that can supplement or replace current traits

## AMS provide udder health indicators

- Electrical conductivity (EC)
  - Standard equipment
  - Quarter level
  
- Online CellCounter (OCC)
  - Available per cow per milking



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### A genetic study of new udder health indicator traits with data from automatic milking systems

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### Data

- EC from 77 herds, 4700 cows, 2.3 mill milkings
- OCC from 24 herds, 1500 cows, 400.000 milkings

# Heritability and genetic correlations

Estimated heritability on diagonal and genetic correlations below diagonal. Standard error in parenthesis

Trait	ECmax	ECmean	InOCC
ECmax	0.23 (0.02)		
ECmean	0.99 (0.002)	0.35 (0.03)	
InOCC	-0.0004 (0.17)	0.04 (0.16)	0.09 (0.03)

- EC traits are heritable
- No genetic correlation between EC and OCC
- Value of EC as indicator for genetic evaluation of udder health is questionable

From Wethal et al. (2020)



# Sensor data can be used for genetic evaluations

## Advantages and new opportunities:

- Automatic, continuous recording
- Objective, frequent measures
- Records already exist for a large proportion of the population
- New traits that can supplement or replace current traits

## Challenges:

- Different systems, providers, versions
  - Data quality
    - Calibration, maintenance procedures, follow up
    - Need validation
  - Access to data
    - «raw data» - summarized data – processed data
    - Insight / documentation of algorithms/variables
  - Data ownership
  - Vast amount of data generated
    - Complex data - Recorded for herd management - How to best utilize it for breeding and genetic evaluation?
  - Routine collection of data to central database
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