



The PigIT Closing Conference

Book of Presentations I

Tuesday, November 13th, 2018, University of Copenhagen

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PigIT report No. 10 • November 2018



This note is also available on www at URL:
<http://www.pigit.net/publications/PigIT-Report10.pdf>

Centre for Herd-oriented Education, Research and Development
Department of Veterinary and Animal Sciences
University of Copenhagen

Contents

1	The Closing Conference	7
1.1	Welcome	7
1.2	Program	8
1.2.1	Arrival and registration	8
1.2.2	Theme 0: PigIT - motivation and core concepts	8
1.2.3	Theme 1: Model based production and welfare monitoring	8
1.2.4	Theme 1 (continued): Model based production and welfare monitoring	8
1.2.5	Theme 2: Model based decision support	8
1.2.6	Theme 3: Where to go from here	9
2	Book of presentations I	10
2.1	Theme 0: PigIT - motivation and core concepts	10
2.1.1	The ideas behind the PigIT project	10
2.2	Theme 1: Model based production and welfare monitoring	23
2.2.1	A dynamic approach to monitoring of growth in finishers	23
2.2.2	Automatic learning and pattern recognition using sensor data	32
2.2.3	Spatial modeling of drinking patterns as a tool for reducing alarms in pig production	42
2.2.4	Behavioral changes preceding tail biting and pen fouling in slaughter pig pens	56
2.2.5	Using machine learning to predict tail-biting, fouling and diarrhea in pigs	68

PigIT partners:



Affiliate partners of PigIT:



Supported by:



Chapter 1

The Closing Conference

1.1 Welcome

It is a great pleasure to welcome our international and Danish connections from science and industry to this event where we share the results and experiences of the seven year research project within the field of precision pig farming.

The **PigIT** project, with the full title *Improving welfare and productivity in growing pigs using advanced ICT methods*, is a strategic research alliance supported by the Danish Council for Strategic Research. It was initiated in 2012 and was, originally, planned to run for five years. It has later been extended by two years.

The partners of the alliance are University of Copenhagen, Aarhus University and the Danish Pig Research Centre. The private companies TNM A/S, Skov A/S and AgroSoft A/S (from 2014) have been affiliated partners.

The scientific objectives of the project have been to develop and improve methods for

1. Model based production and welfare monitoring, and
2. Model based decision support

The themes of the program reflect those two objectives, and the day concludes with perspectives for the future.

1.2 Program

1.2.1 Arrival and registration

The registration desk opens at 9:30 where also a cup of coffee is served.

1.2.2 Theme 0: PigIT - motivation and core concepts

Chair: *Lene Juul Pedersen*

10:00 The ideas behind the PigIT project. *Anders Ringgaard Kristensen*, University of Copenhagen

1.2.3 Theme 1: Model based production and welfare monitoring

Chair: *Lene Juul Pedersen*

10:30 A dynamic approach to monitoring of growth in finishers. *Anna Helena Stygar*, University of Copenhagen (present affiliation: Luke, Finland)

10:50 Automatic learning and pattern recognition using sensor data. *Dan Børge Jensen*, University of Copenhagen

11:20 Coffee break

1.2.4 Theme 1 (continued): Model based production and welfare monitoring

Chair: *Dan Børge Jensen*

11:40 Spatial modeling of drinking patterns as a tool for reducing alarms in pig production. *Katarina Nielsen Dominiak*, University of Copenhagen (present affiliation: SEGES, The Pig Research Centre)

12:10 Behavioral changes preceding tail biting and pen fouling in slaughter pig pens. *Lene Juul Pedersen*, Aarhus University

12:30 Using machine learning to predict tail-biting, fouling and diarrhea in pigs. *Yuvraj Domun*, Harper-Adams, United Kingdom

12:50 Lunch

1.2.5 Theme 2: Model based decision support

Chair: *Anders Ringgaard Kristensen*

13:50 Models and methods for optimization of pig production in a stochastic environment. *Reza Pourmoayed*, Aarhus University (present affiliation: Grundfos)

14:20 From raw data to optimal sow replacement decisions - an integrated solution. *Jeff Hindsborg*, University of Copenhagen

14:40 Coffee break

1.2.6 Theme 3: Where to go from here

Chair: Nils Toft

15:00 The IQinAbox project. *Thomas Nejsum Madsen*, Nejsum Aps

15:20 Ambitions for machine vision. *Dan Børge Jensen*, University of Copenhagen

15:40 Plenary discussion. *All presenters*

16:00 Closing

Chapter 2

Book of presentations I

2.1 Theme 0: PigIT - motivation and core concepts

Chair: Lene Juul Pedersen

2.1.1 The ideas behind the PigIT project

Presentation by Anders Ringgaard Kristensen



The ideas behind the PigIT project

Anders Ringgaard Kristensen



Kick-off seminar, March 8th 2012



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Slide 2



The PigIT alliance 2012 – 2018, who are we?



A strategic research alliance entitled:

- PigIT - Improving welfare and productivity in growing pigs using advanced ICT methods

Strategic vision:

- To create the strongest possible Danish alliance within the field of Precision Livestock Farming (Pigs)
- An internationally leading centre for PLF

Partners:

- University of Copenhagen, Department of Veterinary and Animal Sciences
- Aarhus University, Departments of Animal Science and Economics and Business
- Pig Research Centre

Associate partners:

- TNM A/S
- Skov A/S
- AgroSoft A/S



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Slide 3



Phe PigIT alliance 2012 – 2018, who paid?



Source	DKK ¹	EUR ¹
The Danish Council for Strategic Research	20.0	2.67
University of Copenhagen	3.2	0.43
Aarhus University	2.7	0.36
SEGES, Pig Research Centre	4.4	0.59
Companies	0.8	0.11
Total	31.0	4.13

¹ Millions



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Slide 4



Background



Finisher production in Denmark is under pressure - Increasing export of live piglets to Germany, Poland and other countries

For the Danish finisher production to regain its competitiveness, drastic changes are expected:

- Much larger units
- Less labor input

A consequence will be an increasing number of pigs per employee leaving less time for direct (human) observation of the pigs

Welfare and productivity will be challenged

Smart continuous monitoring by sensors might be one of the ways to go!

Hypothesis: It is the hypothesis that

- a systematic placement of cheap sensors in the production pens
- combined with methodological developments to integrate the information from these sensors will improve the production process.
- The potential benefit will be seen in productivity as well as in the welfare of the animals in the systems.



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Slide 5



The PigIT alliance: Societal objectives



The societal objective of this alliance is to contribute significantly to the competitiveness of the Danish slaughter pig industry while still ensuring a satisfactory level of animal welfare.

This will provide a basis for future commercial products with an international market perspective.

The work packages!

- Growth and feed consumption
- Tail biting
- Diarrhea
- Pen fouling
- Optimal feeding and marketing



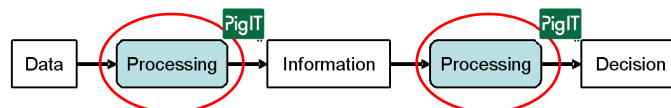
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From data to information to decision



- We refer to a collection (typically in a database) of registrations of the same kind as *data*.
- We don't use data directly for decision making (huge amounts of data).
- Before we can use data we need to reduce it through some kind of processing.
- The resulting information is used for decision making (which again requires processing: optimization).



- Most PLF projects only address the first part of the path from Data to Information (early warning systems, **monitoring**)
- The ambition of PigIT is also to cover the last part of the path from Information to Decision (optimization, **decision support**).



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The PigIT alliance: Scientific objectives

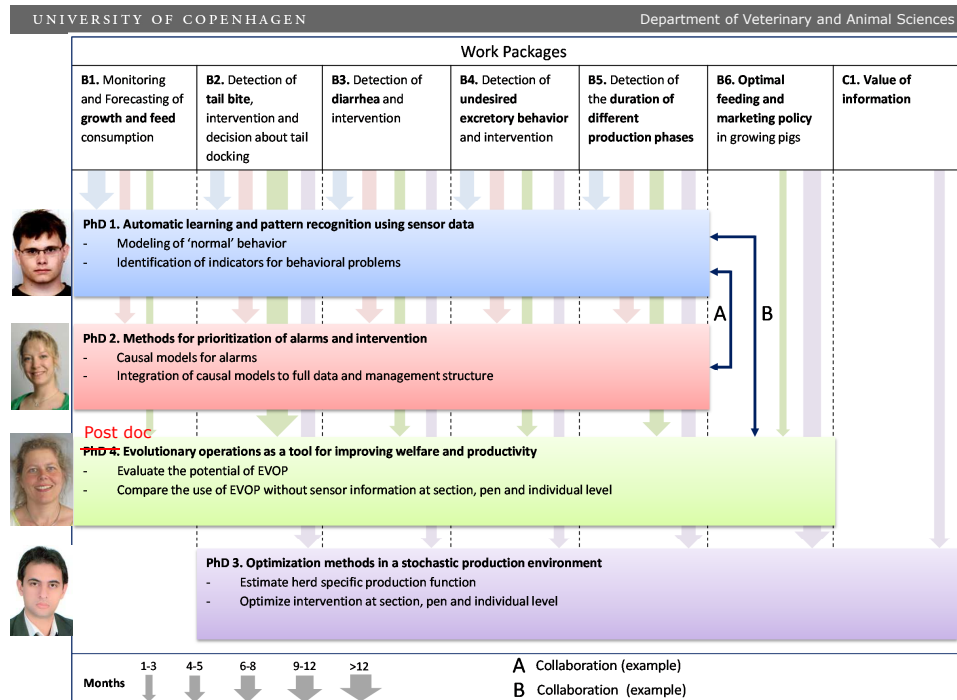
The primary scientific purpose of the alliance is to develop and improve methods for

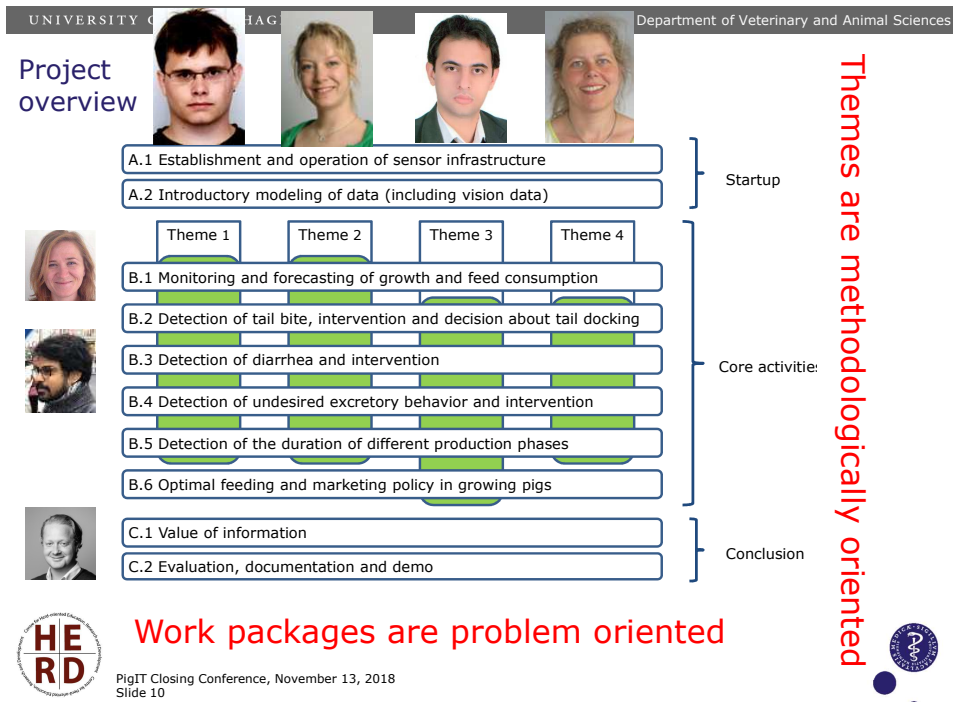
1. Model based production and welfare monitoring by,
 - a) Optimizing the use of automatic measurements based on sensor technology.
 - b) Improving links between data from different animals and different parts of the herd.
 - c) Enhancing the early prediction and identification of reduced welfare and productivity.
2. Model based decision support by,
 - a) Improving integration of data in the decision process by
 - i. learning parameters from data and
 - ii. implementing industrial methods for process design, control, and development.
 - b) Ensuring optimal decisions by integrating indicators of animal welfare and productivity.
 - c) Developing optimization algorithms, which can handle multiple decision criteria and multiple levels.



Slide 8

The methodological themes (PhD projects)





UNIVERSITY OF COPENHAGEN Department of Veterinary and Animal Sciences

Data and herds

PigIT

The PigIT project has collected data in several herds:

- Finishers
 - An experimental herd at Research Centre Foulum, Aarhus University
 - An experimental herd owned by the project partner SEGES, Pig Research Centre, at Grønhøj (Central Jutland)
 - A private herd in Thy (North-Western Jutland)
 - A private herd in Southern Jutland.
 - A private herd with large pens and automatic weighing in Central Jutland
- Weaners
 - An experimental herd at Research Centre Foulum, Aarhus University
 - An experimental herd owned by the project partner SEGES, Pig Research Centre, at Grønhøj (Central Jutland)
 - A private herd in Southern Jutland.

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Slide 11



Types of data in the PigIT project – pen level

Sensor data:

- Temperature
- Water "consumption"

Feeding system (often two-pen level):

- Feed intake by feedstuff per day

Processed video data:

- Weight assessment
- Activity

Unprocessed video data

Farmer: Supplementary observations

- Diarrhea
- Tail biting
- Pen fouling
- Pigs inserted/removed
- Manual weighing (selected pens)



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Slide 12



Types of data in the PigIT project – other levels

Section level

- Climate system:
 - Temperature
 - Humidity

Technician

- Supplementary observations
- Text based logbook

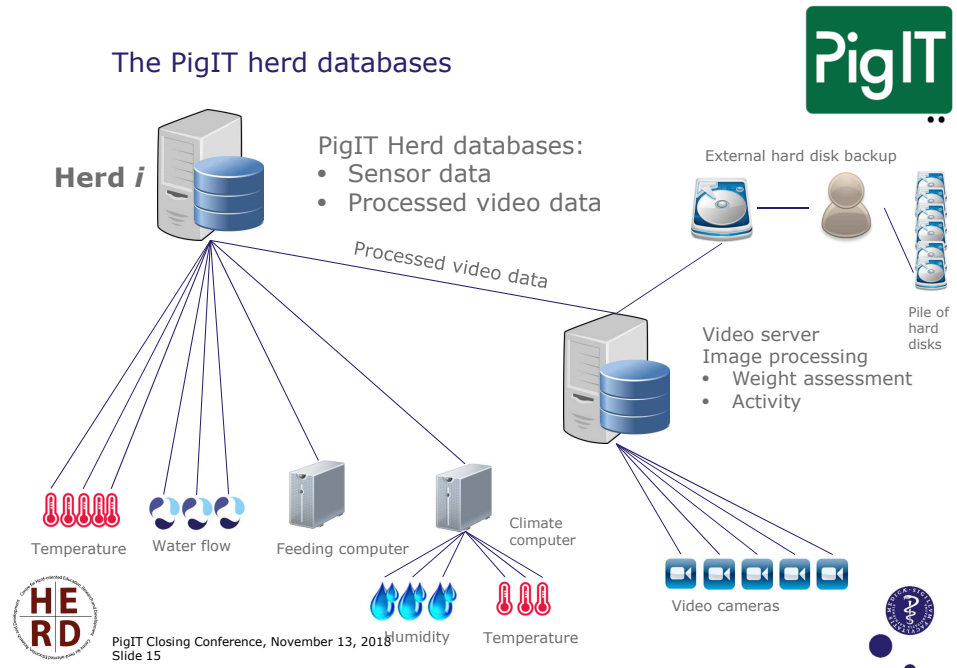
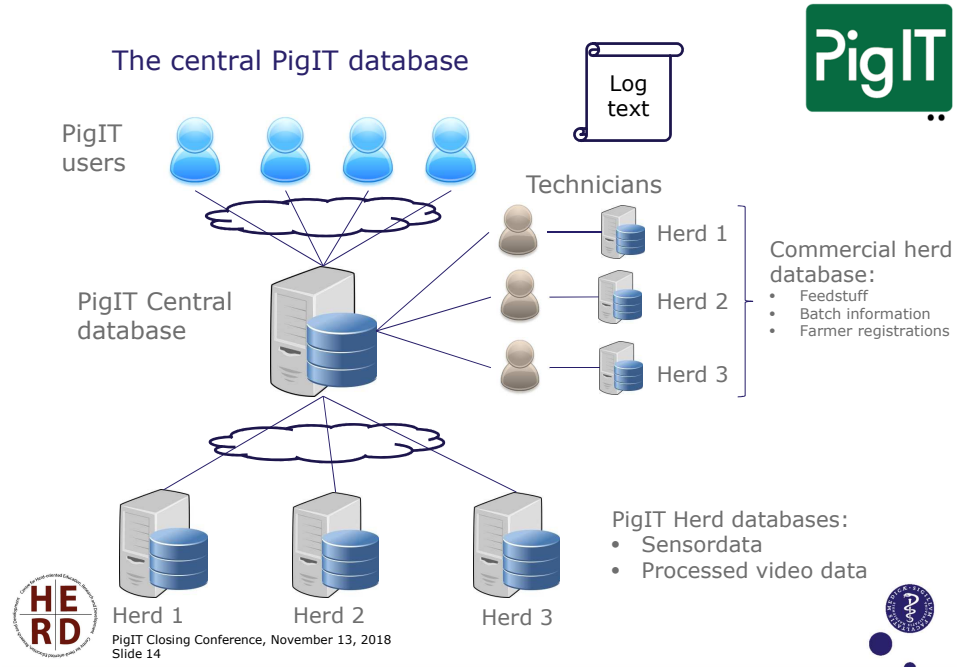
Herd database ("commercial")

- Batch information
- Feedstuff "analysis"



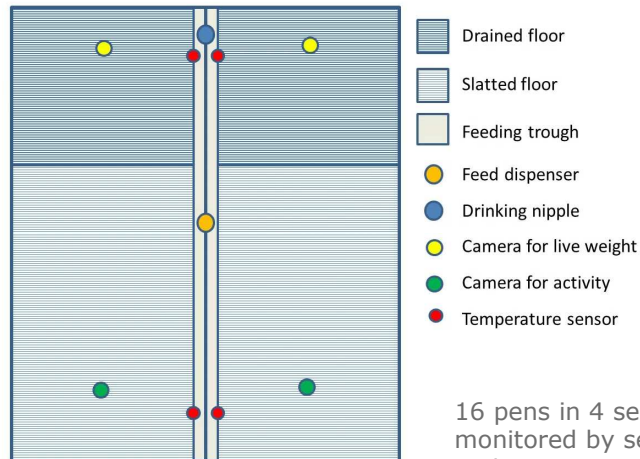
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Sensors as installed in two experimental pens



16 pens in 4 sections are monitored by sensors and cameras



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Dias 16



A dynamic approach to monitoring of growth in finishers



Presentation by Anna Helena Stygar, post doc in PigIT.

Work package:

B.1: Monitoring and forecasting of growth and feed consumption.

Contribution to scientific objectives:

1. Model based production and welfare monitoring by,
 - a) Optimizing the use of automatic measurements based on sensor technology. **Automatic weighing system.**
 - b) Improving links between data from different animals and different parts of the herd. **Herd, batch and pig level**
 - c) Enhancing the early prediction and identification of reduced welfare and productivity. **Reduced productivity**



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Slide 17





Automatic learning and pattern recognition using sensor data



Presentation by Dan Børge Jensen, PhD student, post doc and assistant professor in PigIT.

Work packages:

B.3: Detection of diarrhea and intervention

B.4: Detection of fouling and intervention

C.1: Value of information

Contribution to scientific objectives:

1. Model based production and welfare monitoring by,
 - a) Optimizing the use of automatic measurements based on sensor technology. **Multivariate framework: Water, feed, temperature, humidity**
 - c) Enhancing the early prediction and identification of reduced welfare and productivity. **Diarrhea and fouling**



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Slide 18



Spatial modeling of drinking patterns as a tool for reducing alarms in pig production



Presentation by Katarina Nielsen Dominiak, PhD student and post doc in PigIT.

Work packages:

B.2: Detection of tail biting, interv. and dec. about tail docking

B.3: Detection of diarrhea and intervention

B.4: Detection of fouling and intervention

Contribution to scientific objectives:

1. Model based production and welfare monitoring by,
 - a) Optimizing the use of automatic measurements based on sensor technology. **Water monitoring**
 - b) Improving links between data from different animals and different parts of the herd. **Herd, batch and pen level**
 - c) Enhancing the early prediction and identification of reduced welfare and productivity. **Tail biting, diarrhea and fouling**



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Slide 19





Behavioral changes preceding tail biting and pen fouling in slaughter pig pens



Presentation by Lene Juul Pedersen, professor at AU.

Work packages:

B.2: Detection of tail biting, interv. and dec. about tail docking

B.4: Detection of fouling and intervention

Contribution to scientific objectives:

1. Model based production and welfare monitoring by,
 - c) reduced welfare and productivity. Does behaviour change when tail biting and fouling occur?



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Slide 20



Using machine learning to predict tail-biting, fouling and diarrhea in pigs



Presentation by Yuvraj Domun, Harper-Adams University, visiting PhD student in PigIT

Work packages:

B.2: Detection of tail biting, interv. and dec. about tail docking

B.3: Detection of diarrhea and intervention

B.4: Detection of fouling and intervention

Contribution to scientific objectives:

1. Model based production and welfare monitoring by,
 - a) Optimizing the use of automatic measurements based on sensor technology. Water monitoring, temperature etc
 - c) Enhancing the early prediction and identification of reduced welfare and productivity. Tail biting, diarrhea and fouling



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Slide 21





Models and methods for optimization of pig production in a stochastic environment



Presentation by Reza Pourmoayed, PhD student in PigIT

Work package:

B.6: Optimal feeding and marketing policy in slaughter pigs

Contribution to scientific objectives:

2. Model based decision support by,
 - a) Improving integration of data in the decision process by
 - i. learning parameters from data and **Learning growth and feed intake**
 - b) Ensuring optimal decisions by integrating indicators of animal welfare and productivity. **Productivity**
 - c) Developing optimization algorithms, which can handle multiple decision criteria and multiple levels. **Herd constraints**



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Slide 22



From raw data to optimal sow replacement decisions - an integrated solution



Presentation by Jeff Hindsborg, research assistant in PigIT

Work package:

C.2: Evaluation, documentation and dissemination

Defined deliverables from work package:

- a) Visual prototypes of a monitoring and decision support system. **Not just a prototype but a working software tool due to a separate grant from a levy foundation**
- b) Closing conference where the most important results of the activities are presented – **That's why we are here**



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Slide 23



The future – where to go from here



Two presentations:

1. IQinAbox – presentation by Thomas Nejsum Madsen.
Implementation an commercialization of scientifically based monitoring methods
2. Ambitions for machine vision presented by Dan Børge Jensen, the CYBELE project

Discussion



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Slide 24



Thank you to:



Those who helped us:

- The involved farmers and their staff
- The technicians from SEGES Pig Research Centre
- The technical staff at Aarhus University
- The international Advisory Board

Those who paid:

- The Danish Council for Strategic Research
- SEGES Pig Research Centre
- University of Copenhagen
- Aarhus University



The Danish Council
for Strategic Research

Our affiliate partners:

- TNM A/S
- Skov A/S
- AgroSoft A/S

And last, but not least, to

Those, who did the work:

- The PigIT project group

All of you for listening!



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Slide 25



2.2 Theme 1: Model based production and welfare monitoring


Chair: Lene Juul Pedersen and Dan Børge Jensen

2.2.1 A dynamic approach to monitoring of growth in finishers



Presentation by Anna Helena Stygar


UNIVERSITY OF COPENHAGEN

Department of Veterinary and Animal Sciences



Faculty of Health and Medical Sciences






A dynamic approach to monitoring growth in finishers

Anna H. Stygar^{1,2}

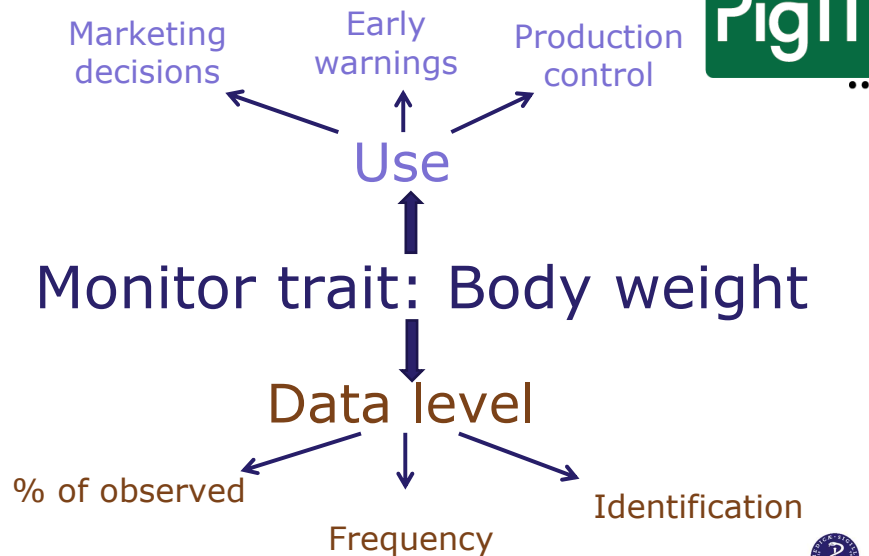
¹Department of Veterinary and Animal Sciences, University of Copenhagen

² Natural Resources Institute Finland (Luke), Bioeconomy and environment, Latokartanonkaari 9, 00790 Helsinki

anna.stygar@luke.fi



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Objectives of this study

- To construct models for growth monitoring and forecasting for finishers using:
 - ❑ the data from traditional BW monitoring based on selected group of pigs (Study 1)
 - ❑ the data from frequent BW monitoring (Study 2)
 - Individually identified pigs (alarms for the whole batch and individual pigs)
 - Unidentified pigs (raising alarms for the whole batch)

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Material and methods



- 2 farms



- Nr of pigs 12,568
- Nr of observations: 397,157

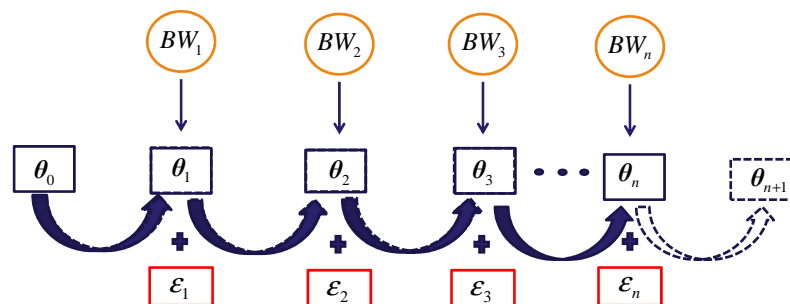
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Material and methods



Dynamic linear model (DLM)



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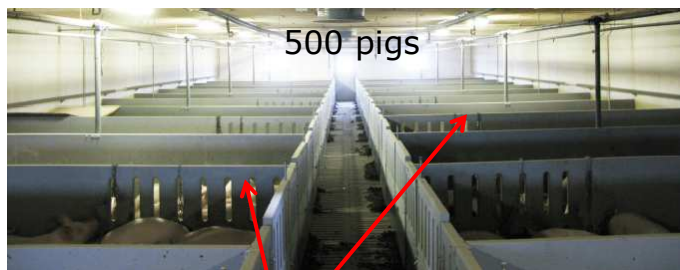


Study 1

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Study 1



15% of pigs in a control group

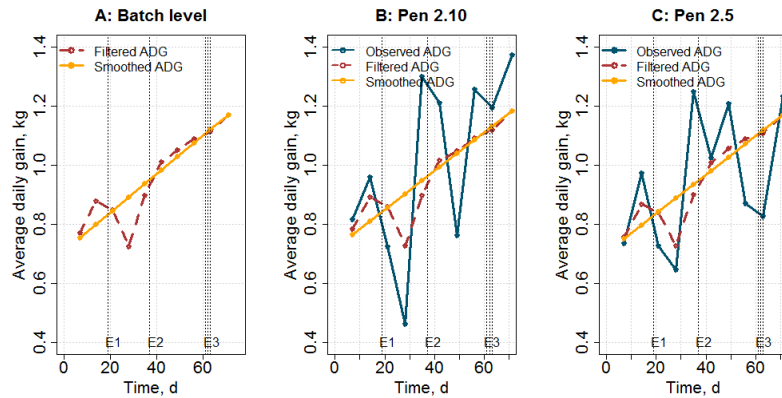
- ~ 600 BW observations
- 1 BW/pig from selected group/week
- **growth control and delivery strategy**

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Study 1 - results

DLM for production control at batch and pen level

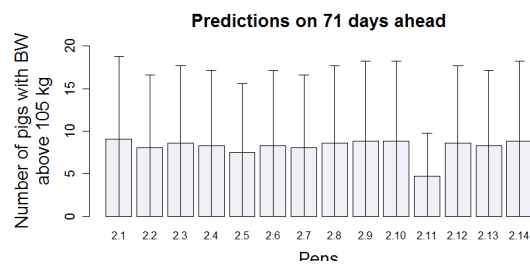


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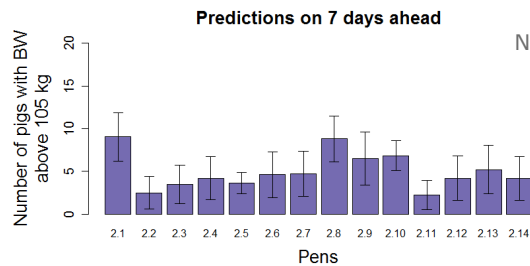
Study 1 - results

DLM for predicting the number of pigs ready to market



Nr of BW observations= 0

MAD=3.6 pig
CV = 102%



Nr of BW observations ~ 1100

MAD=1.6 pig
CV = 24%





Study 2

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Study 2



- 40,000 -60,000 BW observations
- 3.1 -4.5 BW/pig/day
- currently provides delivery strategy



Growth alarms for identified and unidentified pigs?

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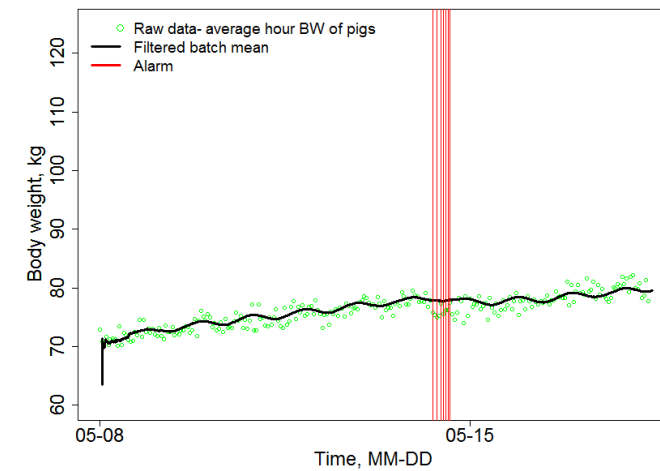


Study 2 - results

DLM with diurnal growth pattern



A: Alarms for marked pigs



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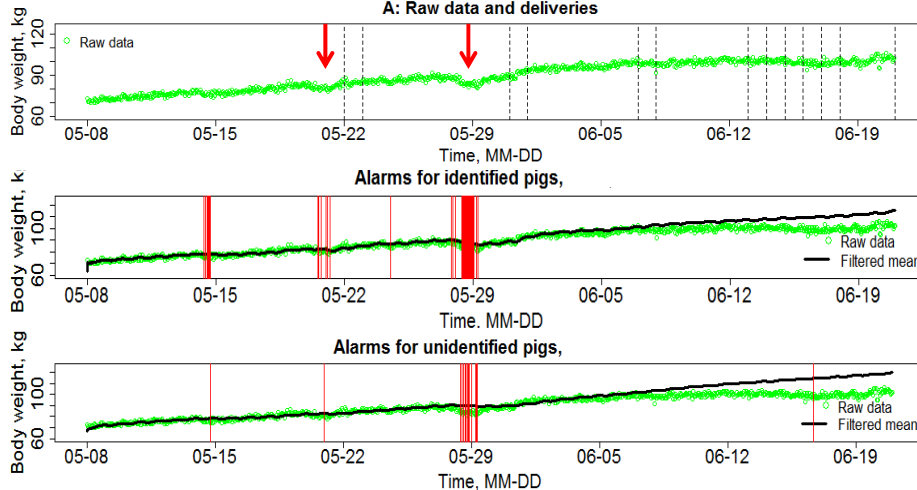


Study 2 - results

Alarms for identified and unidentified pigs



A: Raw data and deliveries

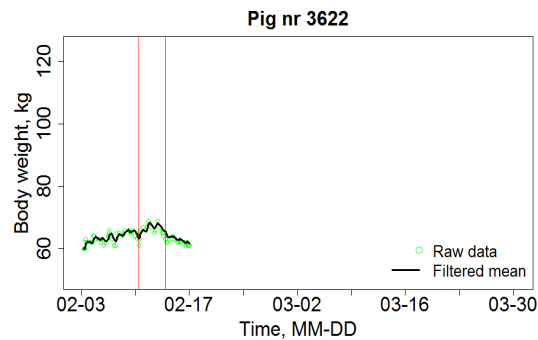


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Study 2 - results

Alarms for individual pigs



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Conclusions



- Few new solutions for BW monitoring (3 models):
 - Production control
 - Predicting number of pigs ready to market
 - Frequent monitoring
- Flexible tools – using different level of information
- New view on BW in pigs for frequent monitoring – diurnal pattern
- Batch level alarms on growth are possible with unidentified pigs

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Perspectives for future research and for pig farming



- Economic analyses on value of information
- Combining frequent BW observations with other sensor data
- Testing and calibrating constructed models at a farm level
- Possible application for assessing effect of an intervention (e.g. change in feeding strategy, new vaccinations, modified herd management)

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Study 1

Monitoring growth in finishers by weighing selected groups of pigs – A dynamic approach1 [doi: 10.2527/jas.2015-9977](https://doi.org/10.2527/jas.2015-9977)

Study 2

Analyses of body weight patterns in growing pigs: a new view on body weight in pigs for frequent monitoring
[doi: 10.1017/S1751731117001690](https://doi.org/10.1017/S1751731117001690)

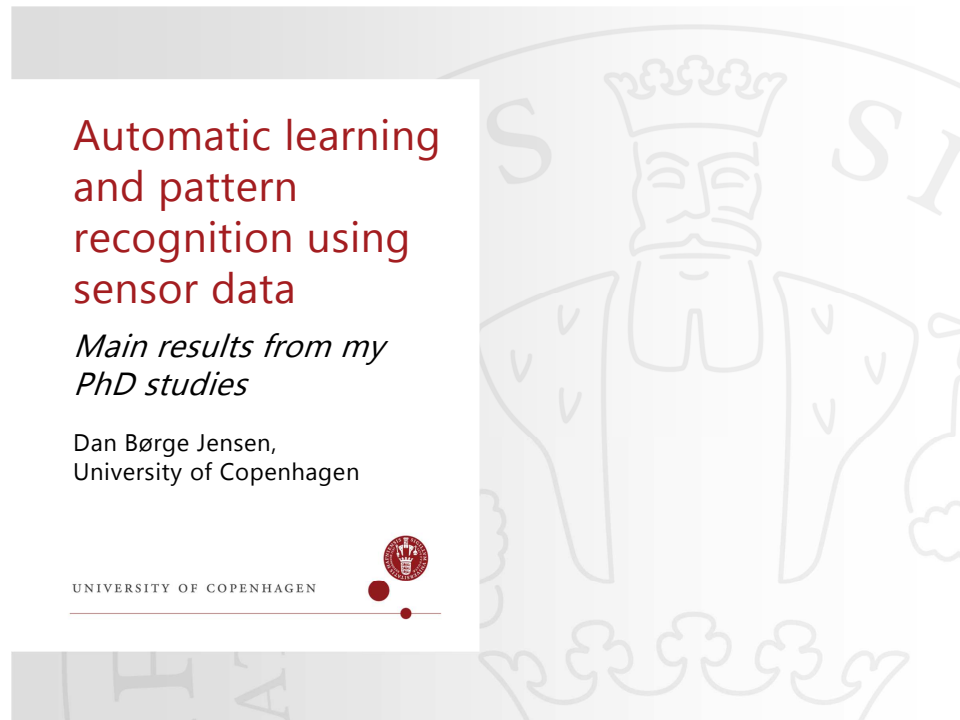
Detecting abnormalities in pigs' growth – A dynamic linear model with diurnal growth pattern for identified and unidentified pigs
[doi: 10.1016/j.compag.2018.10.004](https://doi.org/10.1016/j.compag.2018.10.004)

Thanks to:
PigIT
The Danish Council for Strategic Research
Anonymous farmers



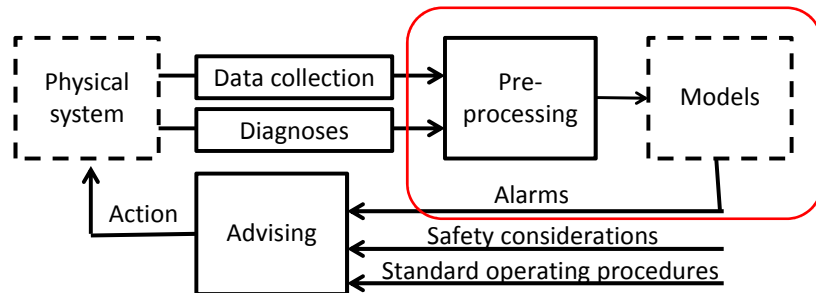
2.2.2 Automatic learning and pattern recognition using sensor data

Presentation by Dan Børge Jensen



Introduction (1)

• The model-based monitoring system



Introduction (2)

<p>ELSEVIER</p> <p>The effect of daily weight of grower</p> <p>Dan B. Jensen</p> <p>14800 - Growth for 1 University of Copenhagen Veterinary Institute</p> <p>ARTICLE INFO</p> <p>Article history: Received 13 March 2017 Received in revised form 22 June 2018 Accepted 10 June 2018</p> <p>Keywords: Pig Pig production Weight gain Feed efficiency</p> <p>1. Introduction</p> <p>Control of pig growth is currently at an increasing rate, due to the demand for leaner meat (Fitzpatrick, 2015). However, the ability to predict the growth of individual pigs is still limited, which is the one of the main challenges in pig production.</p>	<p>ELSEVIER</p> <p>Bayesian inference in dynamic linear models</p> <p>Dan B. Jensen,^a Dorte Borge Jensen,^a Department of Large Animal Medicine, 20 Department of Animal Health</p> <p>ARTICLE INFO</p> <p>Article history: Received 20 April 2018 Received in revised form 22 August 2018 Accepted 22 December 2018</p> <p>Keywords: Bayesian inference Dynamic linear models Pig production Pig</p> <p>1. Introduction</p> <p>Although it is not a new type of data, it is used in systems for pig production. For example, we have seen a growing interest in early detection of disease (Jensen, 2015). This is known as a key factor in the pig industry, where the pig is the main asset.</p>	<p>PLOS</p> <p>A multivariate analysis of pen fouling in a pig production system</p> <p>Dan Borge Jensen</p> <p>ARTICLE INFO</p> <p>Article history: Received 20 April 2018 Received in revised form 22 August 2018 Accepted 22 December 2018</p> <p>Keywords: Dynamic linear models Pig production Pig</p> <p>1. Introduction</p> <p>Although it is not a new type of data, it is used in systems for pig production. For example, we have seen a growing interest in early detection of disease (Jensen, 2015). This is known as a key factor in the pig industry, where the pig is the main asset.</p>	<p>J. Dairy Sci. 101(233-245) https://doi.org/10.3168/jds.2017-12828 © American Dairy Science Association, 2018.</p> <p>Describing temporal variation in reticulorumenal pH using continuous monitoring data</p> <p>M. J. Denwood,^a J. L. Klean,^a D. B. Jensen,^a and N. H. Johnson^{b,c}</p> <p>^aDepartment of Veterinary and Animal Sciences, University of Copenhagen, 1870 Frederiksberg C, Denmark ^bCoventry, California, 94622, Germany ^cDepartment of Population Health and Reproductive Science, University of Cambridge, Cambridge, CB1 1TA, United Kingdom</p> <p>ABSTRACT</p> <p>Reticulorumenal pH has been linked to subclinical disease in dairy cattle, leading to considerable interest in identifying pH observations below a given threshold. The relatively recent availability of continuously monitored data from pH boluses gives new opportunities for characterizing the normal patterns of pH over time and distinguishing these from abnormal patterns using more sensitive and specific methods than simple thresholds. We fitted a series of statistical models to continuously monitored data from 93 animals on 13 farms to characterize normal variation within and between animals. We used a subset of the data to make deviations from the normal pattern to the predictability of 21 dairy cows from a single herd. Our findings show substantial variation in pH characteristics between animals, although animals within the same farm tended to show more consistent patterns. There was strong evidence for a predictable diurnal variation in pH which could be explained using a simple statistical model. For the 21 animals with available production information, there was also a strong association between productivity (as measured by both milk yield and dry matter intake) and deviation from the expected diurnal pattern of pH 2 d before the productivity observation. In contrast, there was no association between productivity and the occurrence of observations below a threshold.</p> <p>Key words: reticulorumenal pH, acidosis, rumen, monitoring data, statistical model</p> <p>INTRODUCTION</p> <p>Assessment of reticulorumenal pH in cattle has been used as an indicator of an excessive intake of soluble carbohydrates or a shortage of physically effective fiber and the consequent predisposition to a range of health problems, including ruminitis, liver abscess, left displaced abomasum, diarrhea, laminitis, and poor milk production (Deeken and Smith, 1987; Nordlund and Garrett, 1997; Garrett et al., 1998; Palmer et al., 2008). Evaluation of reticulorumenal pH is therefore of interest, with most clinicians using a single observation to define the status (Nordlund et al., 1995; Nordlund and Garrett, 1997). It is now possible to continuously monitor pH values using rumen-sensing data, but this generates a large volume of data that can be very challenging to interpret. As a result, researchers using continuous pH monitoring techniques to investigate diet and substrate have most often used average values or threshold approaches for evaluating reticulorumenal pH. For example, Kheifets et al. (2011) used the overall mean pH for an individual animal over a period of time and the number of minutes that the pH was seen to be below 5.6. We aim to define and diagnose abnormal reticulorumenal pH by identifying deviations from normal pH patterns rather than the detection of</p>
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+ 4 conference papers
+ 2 papers in review
+ some still-unfinished work

FINDINGS

Paper 1

Pen level environment matters!

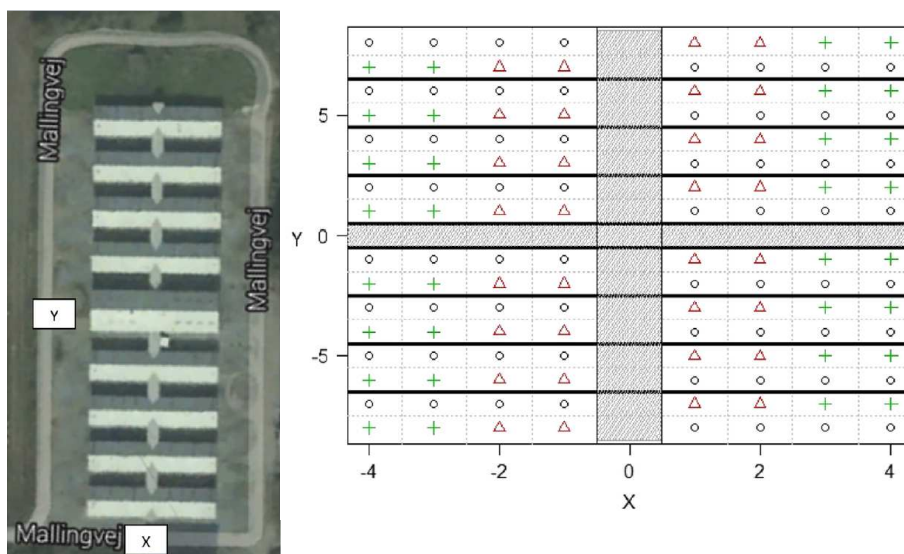
Published in Livestock Science:

Jensen, D.B., Toft, N. & Cornou, C., 2014.

The effect of wind shielding and pen position on the average daily weight gain and feed conversion rate of grower/finisher pigs.

Livestock Science, (167), pp.1–9.

Pen level environment matters! (1)



Pen level environment matters! (2)



Automatic monitoring of
feed consumption



Weekly live weight measurements

Pen level environment matters! (3)

Results:

- **Shielding**

- Unshielded pigs grow more slowly during the winter
- Larger-than-average pigs grow faster without shielding (except in the winter)
- The feed conversion rate was not affected

- **Distance from section entrance**

- Larger-than-average pigs grow faster when placed away from the section entrance
- Average daily feed intake was larger for pigs placed away from the entrance
- The feed conversion rate was not affected

FINDINGS

Paper 2

Pen-level temperature predicts undesired events (pen fouling or diarrhoea)

Published in Livestock Science:

Jensen, D.B. & Kristensen, A.R., 2016.

Temperature as a predictor of fouling and diarrhea in slaughter pigs.

Livestock Science, 183, pp.1–3.

Temperature predicts undesired events (1)

• Assumption:

- Probability of **diarrhea and/or pen fouling** depends on daily **temperature summary** variables, such as:

- Min. temperature
- Max. temperature
- Greatest temperature increase
- Greatest temperature decrease

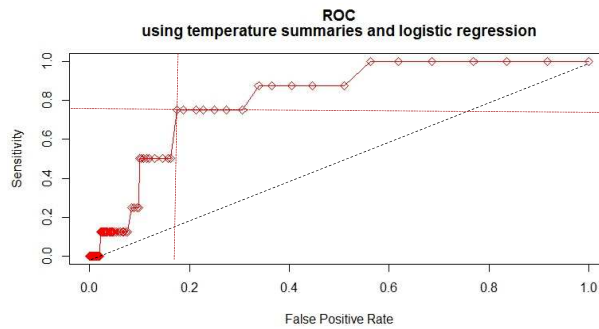
8 variables in total

Variables	Estimate	Std. error	p-Value
(Intercept)	– 12.78	3.53	0.0003
Maximum temperature, drinking nipple	– 0.89	0.35	0.011
Minimum temperature, drinking nipple	1.24	0.43	0.004
Greatest temperature decrease, drinking nipple	– 2.13	0.87	0.014
Greatest temperature increase, corridor	2.70	0.90	0.003
Greatest temperature decrease, corridor	1.53	0.89	0.086

Logistic regression model

Temperature predicts undesired events (2)

- **Performance:**



AUC:

Diarrhea or pen fouling: 0.80

Pen-level temperature is informative!

FINDINGS

Paper 3

**The DLM/Cholesky method can be used for
indiscriminant event detection**

Jensen, D.B., Toft, N., Kristensen, A.R., 2017. *A multivariate dynamic linear model for early warnings of diarrhea and pen fouling in slaughter pigs*. Computers and Electronics in Agriculture, 135, pp. 51–62.

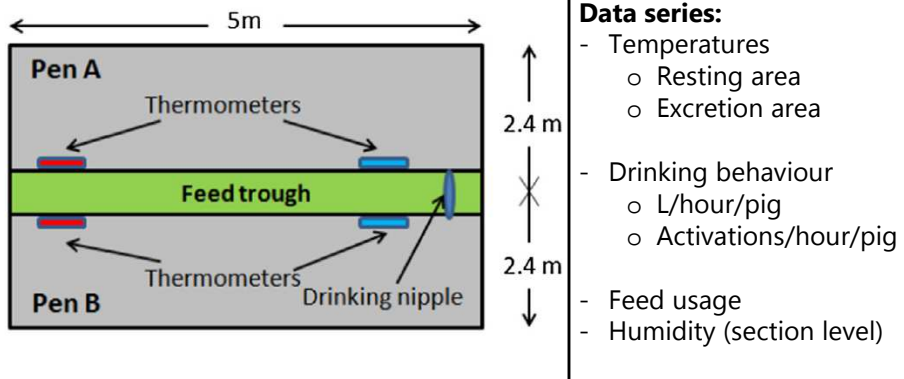
The DLM/Cholesky method (1)

Data source:

Commercial Danish pig farm

12 pens (6 double pens), 3 grower/finisher periods

Events: Diarrhea (N=11) and pen fouling (N=13)



The DLM/Cholesky method (2)

- the Dynamic Linear Model

• Structure:

• Observation equation

$$Y_t = F_t' \theta_t + v_t, \quad v_t \sim N(\underline{0}, V)$$

• System equation

$$\theta_t = G_t \theta_{t-1} + w_t, \quad w_t \sim N(\underline{0}, W)$$

θ_t	Parameter vector
F_t	Design matrix
G_t	System matrix
V	Observational variance
W	System variance

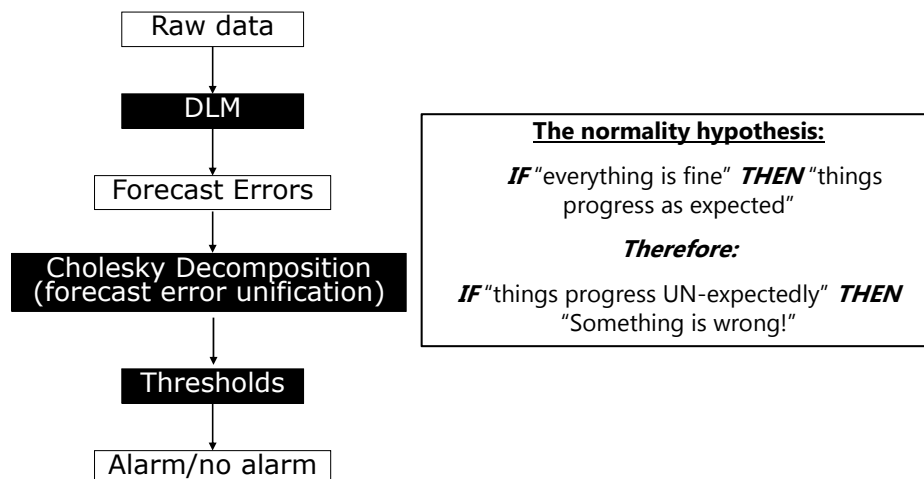
• Usefulness:

- Monitoring of (production) systems over time

• Features:

- Provides one-step-ahead forecasts, including estimated forecast variance
- Dynamic, *i.e.* Adaptive

The DLM/Cholesky method (3)



The DLM/Cholesky method (4)

- **Performance – which variables contribute?**

Omitted variables	AUC
None	0.84
Mean feed usage	0.84
Humidity	0.83
Temperature, excretion area	0.84
Temperature, resting area	0.84
Temperatures, excretion & resting area	0.84
Water flow	0.80
Drinking bouts frequency	0.78
Water flow & drinking bouts frequency	0.76

How many consecutive alarms?
 (0-25 hours → ROC curve)

The DLM/Cholesky method (5)

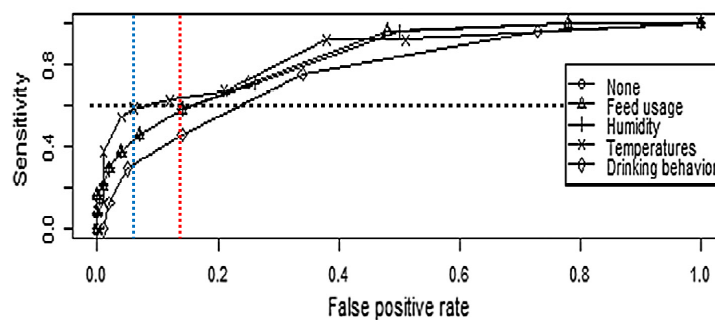
- Performance – which variables contribute?

Included variables	AUC
All	0.84
Feed amount and Humidity	0.52
Temperatures, excretion & resting area	0.75
Drinking behavior	0.84

Unlike
Paper 2

How many consecutive alarms?
(0-25 hours → ROC curve)

The DLM/Cholesky method (6)



Nothing omitted: AUC = 0.84 Se = 60 → Sp = 0.85

Temperature omitted: AUC = 0.84 Se = 60 → Sp = 0.95

How many consecutive alarms?
(0-25 hours → ROC curve)

Conclusions



Main conclusions

Precision data are **useful** for detecting problems relating to health and welfare in animal production

Dynamic linear models can be used to meaningfully combine **multiple diverse** sources of information

- Can determine relative information values!

Temperature data and **drinking behavior** data are both **informative**, with respect to undesired events

- However, they may be best utilized in different ways!

There is further **potential for event-specific** predictions

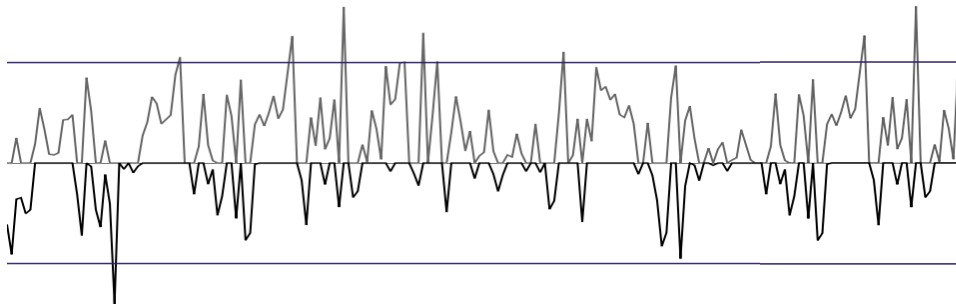
2.2.3 Spatial modeling of drinking patterns as a tool for reducing alarms in pig production

Presentation by Katarina Nielsen Dominiak

Spatial modeling of drinking patterns as a tool for reducing alarms in pig production

Katarina Nielsen Dominiak

PigIT Closing Conference • 13-11-2018



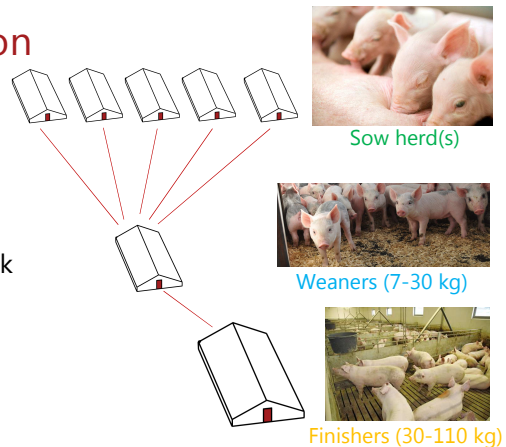
Outline

- Background and basic ideas behind my study
- Objective of my study
- Materials and methods
- Main conclusions
- Future perspectives for research and for pig farming

Modern pig production is centralized

More animals per herd → less time per animal during daily check

Early signs of outbreaks might be missed during a busy day



Sensor-based monitoring around the clock →

Real-time notifications of early signs of outbreaks →

Timely interventions may prevent actual outbreaks

A general challenge is:

Too many false alarms

- Costly and time consuming
- Diminish the trust in the detection system
- Devalue the managerial value

Literature review:

Only three papers focus on methods for reducing false alarms

Dominiak and Kristensen (2018)

Sensor-based detection models - threshold



Low threshold →

- Few sick animals missed
- Many alarms for healthy animals

One alarm may reflect multiple conditions



General wellbeing⁽¹⁾

Tail bite?⁽²⁾

Stress behaviour?



Respiratory diseases?

Other causes?



Change in feed composition?

⁽¹⁾ Madsen (2005), Dominiak (2017)

⁽²⁾ Dominiak et al (2018)

Outline

- Background and basic ideas behind my study
- Objective of my study
- Materials and methods
- Main conclusions
- Future perspectives for research and for pig farming

Objective of my study

To develop and evaluate a decision support system which

- Categorize
 - Sort
 - Prioritize
 - Rank
- the generated alarms

Spatial Decision Support System

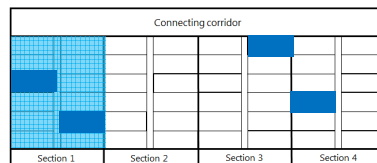
- Area-specific alarms system
- Based solely on data from affordable water sensors (flow meters ~ € 45)

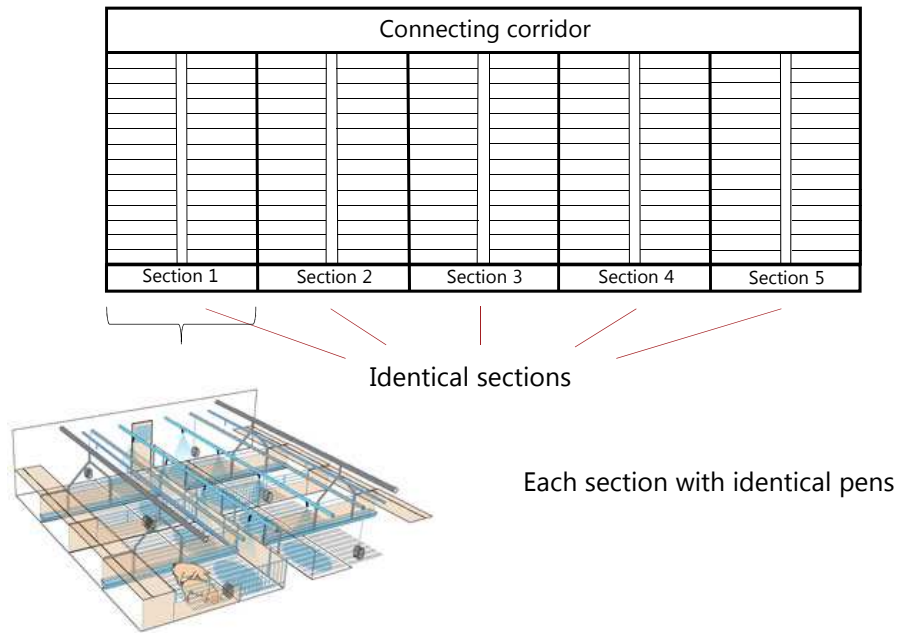


Predicts outbreaks of tail biting, diarrhea or pen fouling

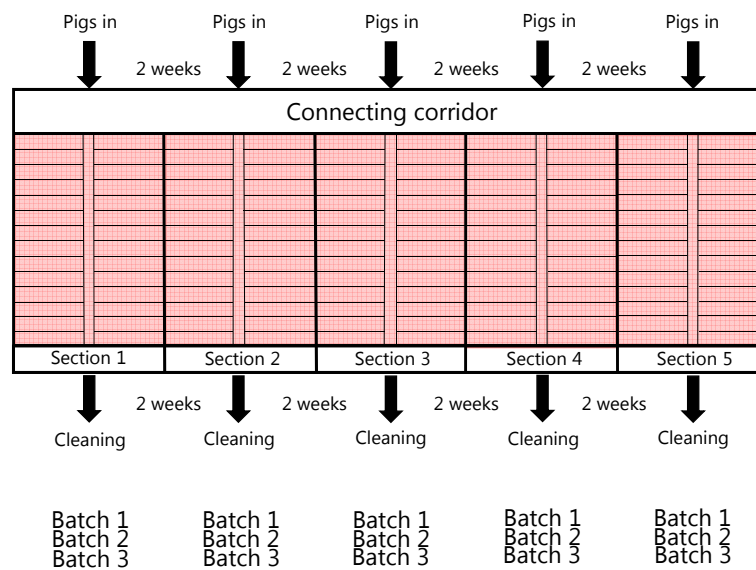
Points out specific pens or sections with abnormal drinking patterns leading to outbreaks

The pointed areas are seen as FOCUS AREAS for **management**





All-In-All-Out bio-security strategy improve animal health



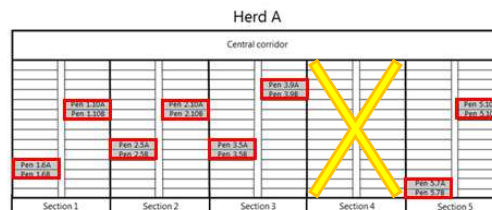
Outline

- Background and basic ideas behind my study
- Objective of my study
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- Main conclusions
- Future perspectives for research and for pig farming

Materials

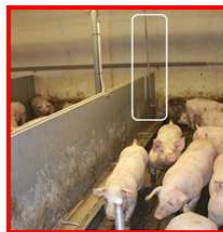
Herd A

- Finisher herd (30-110 kg)
- 4 sections – randomly chosen
- 8 sensors, each monitoring two pens (36 pigs)
- 7 batches (05/2014 – 03/2016)



Herd B

- Weaner herd (7-30 kg)
- 4 sections
- 16 sensors, each monitoring one pen (15 pigs)
- 13 batches (10/2014 – 12/2016)



49

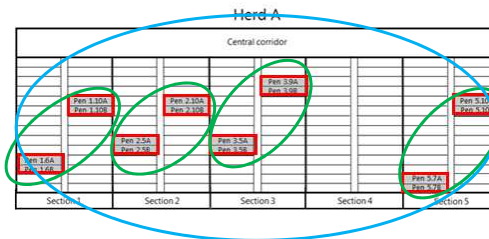
Method

A *Dynamic Linear Model* (DLM) aims to predict the next observation based on all prior observations.

The DLM learns the pattern of data from each sensor and predicts the next observation in each pattern

All new observations are added to the prior knowledge

Model is reset between batches

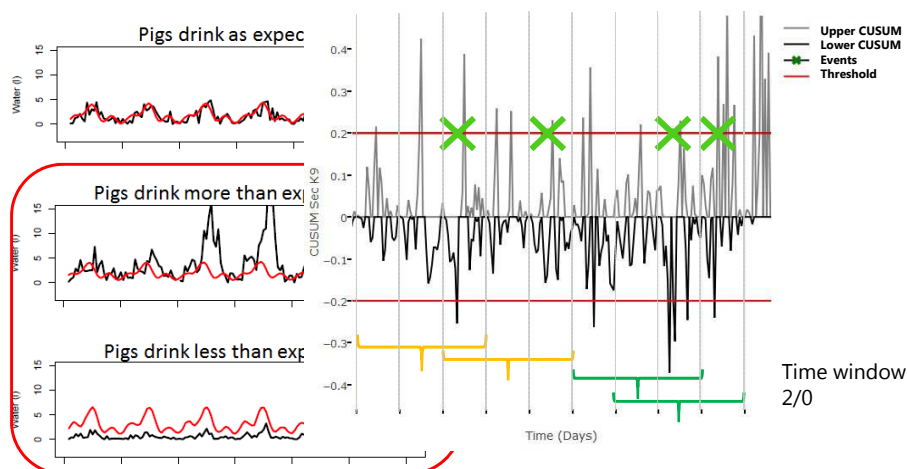


Interactions between drinking patterns are incorporated through variance components in the DLM:

- Pens within the same section are assumed correlated
- Sections within the same herd are assumed correlated

Method

- Raw data, actual drinking pattern
- Fitted, expected drinking pattern
- Cumulated sum of forecast errors in Tabular Cusum
- Forecast errors

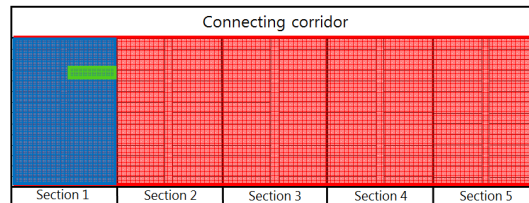


A few results

Unwanted events in any
pen within the herd
AUC (48 hours) = 0.92

Unwanted events in a specific
section in the herd
AUC (48 hours) = 0.87

Unwanted events in a specific
pen in the herd
AUC (48 hours) = 0.87



Herd

Central water supply
Central feeding system
Central electricity

Section

Pigs of same age
Feed mixture
Climate control

Pen

Pigs of same gender/size

Outline

- Background and basic ideas behind my study
- Objective of my study
- Materials and methods
- Main conclusions
- Future perspectives for research and for pig farming

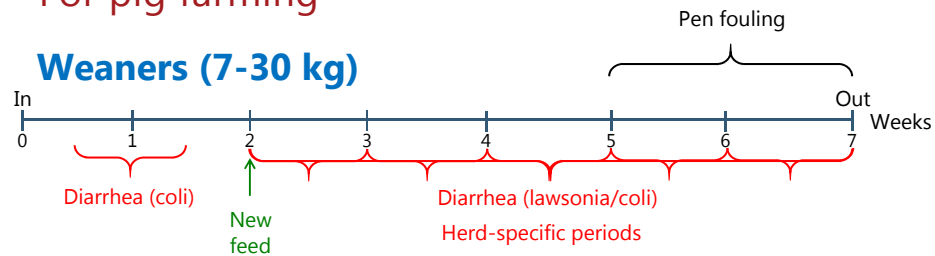
- Alarms for specific pens and sections point out high-risk areas with high accuracy
- Changes in drinking patterns tends to be related to multiple causes
- Productivity and animal welfare can be improved through automatic monitoring of water data

Outline

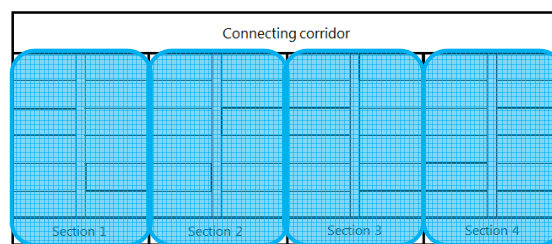
- Background and basic ideas behind my study
- Objective of my study
- Materials and methods
- Main conclusions
- Future perspectives for research and for pig farming

For pig farming

Weaners (7-30 kg)



Alarms for specific pens can be added as additional information



Time after insertion

0-1 week 2-3 weeks 4-5 weeks 6-7 weeks

Management

- Climate
 - Ventilation settings/draft
 - Temperature settings
- Feeding system
 - Change in feed composition
- Water system
 - Blocked drinking nipple
 - Leaking drinking nipple
 - Sufficient water supply?



Reflected in Water consumption

How to respond to area-specific alarms

- Check smart-phone interface
- Go to specific areas where alarms are generated
- If high risk of known condition – choose preventive intervention
- Otherwise check:
 - Climate/ventilation
 - Feeding system
 - Water system



For research



- Include herd-specific knowledge
 - High-risk periods
 - Age and health status of pigs in a section
- Differentiate between higher and lower water consumption

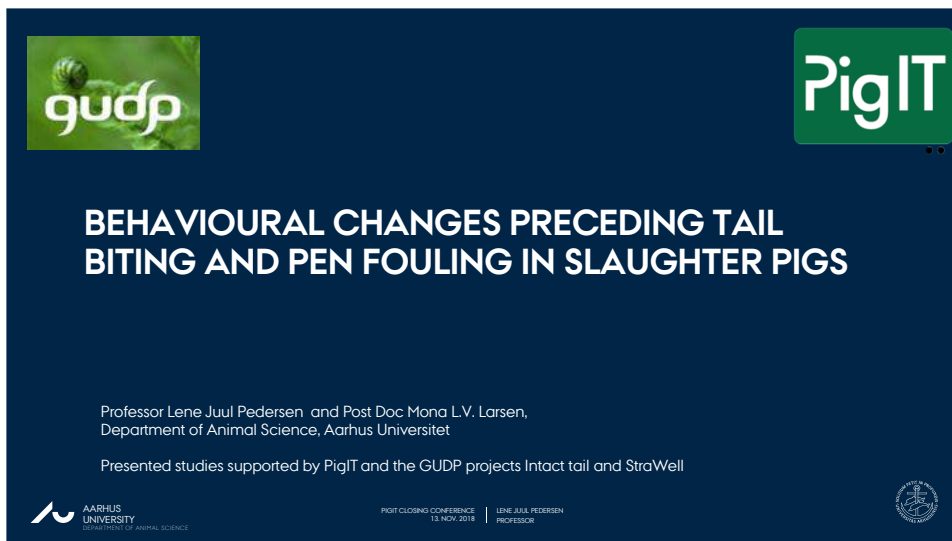
THANK YOU



This research was funded by the Danish Council for Strategic Research (The PigIT project, Grant number 11-116191)

2.2.4 Behavioral changes preceding tail biting and pen fouling in slaughter pig pens

Presentation by Lene Juul Pedersen



gudp

PigIT

BEHAVIOURAL CHANGES PRECEDING TAIL BITING AND PEN FOULING IN SLAUGHTER PIGS


Professor Lene Juul Pedersen and Post Doc Mona L.V. Larsen,
Department of Animal Science, Aarhus Universitet

Presented studies supported by PigIT and the GUDP projects Intact tail and StraWell

AARHUS UNIVERSITY
DEPARTMENT OF ANIMAL SCIENCE

PigIT CLOSING CONFERENCE
13. NOV. 2016

LENE JUUL PEDERSEN
PROFESSOR



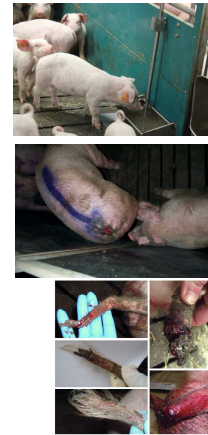
BEHAVIOURAL PROBLEMS IN PIG PENS

Tail Biting:

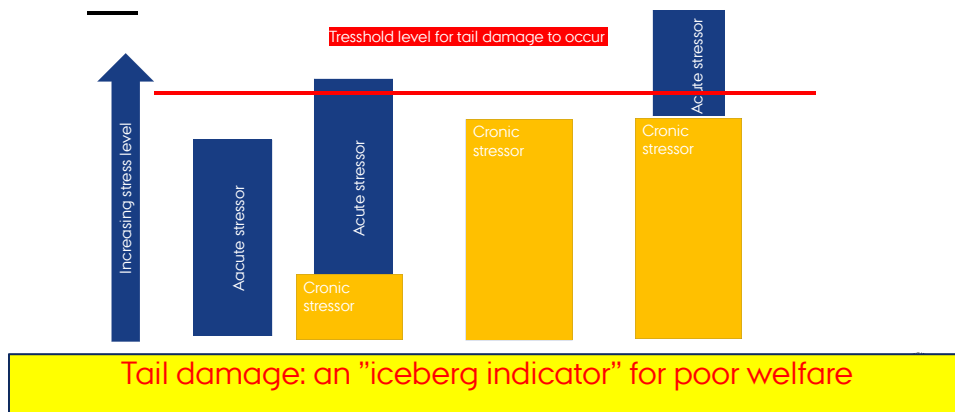
Redicreted explorative behaviour or sudden forceful bites

Occurs as a result of increased stress => stress increases the motivation for explorative behaviour

Nothing to explore, behaviour is redirected towards other pigs tails => tail biting and tail damage



MANY RISK FACTORS ADD UP STRESS (VALROS, 2018)



BEHAVIOURAL PROBLEMS IN PIG PENS

Pen fouling:

- When pigs starts dunging in the resting area
- Results in poor hygiene, increased risk of disease, increased workload
- Risk factors: Inadequate temperatures and high stocking density- indicator of stressfull conditions
- Easily seen in pens with partly solid floor
 - Although pigs in slatted floor pens may be equally affected



WHY EARLY WARNING ?

- Difficult to eliminate all risk factors – eg. changes in temperature
- Useful to develop an early warning system
- Our task was to investigate:
 - Behaviours indicative of increased stress prior to tail damage?
 - Behaviours/thermal patterns indicating pen fouling will soon develop ?
 - Warning => intervention => negative events can be avoided

DATA COLLECTED IN EXPERIMENTAL HERD

112 pens with slaughter pigs (4 bcts)

Daily scoring of pen fouling

Pen fouling: >50 % of solid floor wet with faces and/or urine

Careful inspection of each pig tail 3 times weekly

Tail damage pen: At least ONE pig with bleeding tail



Tails: docked vs. undocked

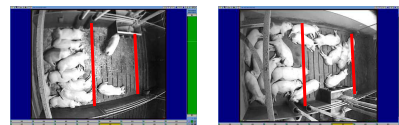
Straw: No straw vs. 150 g straw per pig per day

Space/group size: 0.73 m²/pig (18 pigs)
vs 1.21 m²/pig (11 pigs)

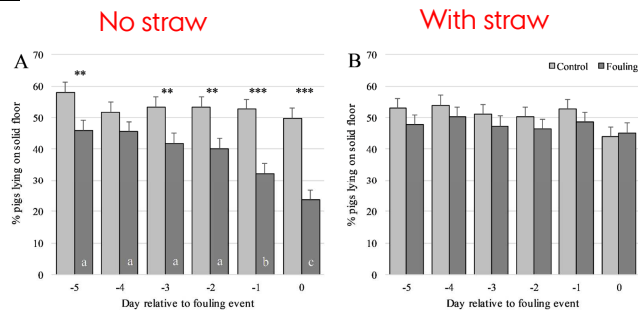


BEHAVIOUR PRIOR TO PEN FOULING

- Pigs lying pattern day -5 to 0 before pen fouling occurred observed from video
- Paired control pens and outbreak pens on same days
- Did lying pattern differ between control and pen fouling pens ?

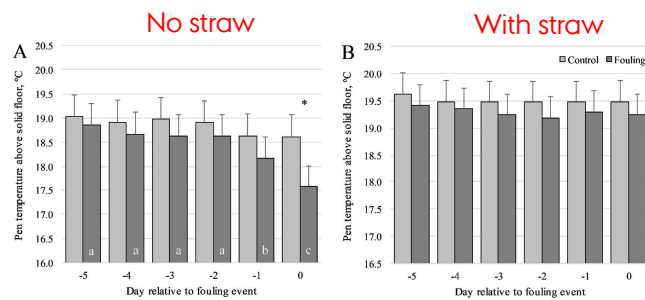


RESULTS- LYING PATTERN



Clear decrease in % pigs lying on solid floor prior to pen fouling
NOT seen in pens with straw

RESULTS- PEN TEMPERATURE SOLID FLOOR



Pen temperature on solid floor reflected lying pattern

PEN FOULING CONCLUSION

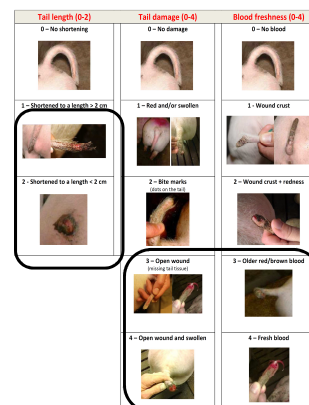
Lying in resting area clearly declined prior to fouling in no-straw pens

Temperature reflected lying pattern – can it be used as early indicator of changed lying pattern ?

Lying in resting area did not decline in straw pens – despite higher temperature - Why ?

BEHAVIOUR PRIOR TO TAIL DAMAGE

- Behavioural changes occur at pen level prior to tail damage ?
- Paired control pens and outbreak pens on same days
- Which behaviours differed between control pens and tail damage pens ?



FROM LITTERATURE: WHICH CHANGES WERE EXPECTED ?

(LARSEN ET AL 2016 REVIEW)

Days preceding outbreak:

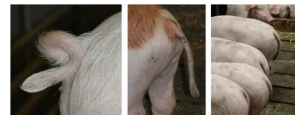
Tail posture: less curly tails

Activity: increased

Exploration: increased

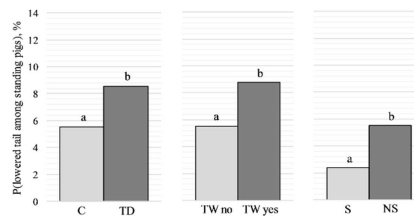


CURLY TAILS



Hanging tails day -3 til -1:

- More in pens developing tail damage day 0
- More in pigs with minor tail wounds
 - Indicates ongoing tail biting behaviour
- More in pens without straw



Larsen et al 2018a

Hanging tails: signal of increased risk of tail damage

ACTIVITY PRECEDING TAIL DAMAGE



Observation of activity day -7 to -1 from video:

Higher activity in pens developing tail damage (TD)

Changes develops PRIOR to day -7

Use of continuous measured activity can increase knowledge

- Automation using eg image analysis needed

Afternoon	P(active), %	
	NON-TD	TD
Day-6 to -7	17.37	21.20
Day-5 to -4	15.06	22.19
Day-3 to -1	14.48	21.49

Larsen et al 2018b



EXPLORATION TOWARD WOODEN BEAMS

Observation of behaviour directed wooden beams day -7 to -1

Less exploration in pens developing tail damage

Differences only present in pens with undocked pigs



Larsen et al 2018b



SUMMARY

Behaviour prior to tail damage do differ from that seen in control pens

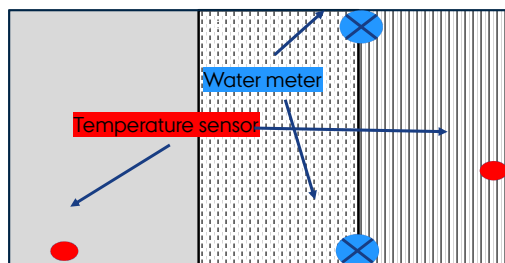
Changes occur prior to day -3/-7 or are permanently present

Can behavioural changes be reflected in automated measures using sensors ?

CURRENTLY BEING INVESTIGATED

Water and temperature sensors in each pen

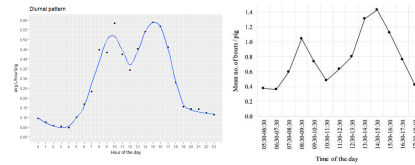
Temperatur and humidity in each section



INFORMATION FROM SENSOR DATA

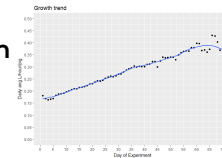
Step 1: Modeling of normal pattern:

- Cyclic diurnal pattern in water consumption:
 - Two peaks : morning and afternoon
 - Similar to diurnal pattern in explorative behaviour



- Increased water consumption with growth:

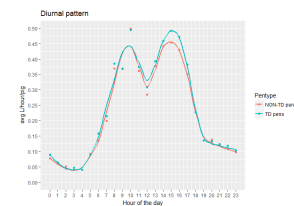
- As pigs grow they drink more



INFORMATION FROM SENSOR DATA

Step 2: Changes prior to tail damage

- Ongoing work
- Extract features of the diurnal pattern and growth trend
- Do these features differ between control and tail damage pens prior to tail damage ?
- Are the features affected by presence of stressor/risk factors (stocking density and straw allocation) ?



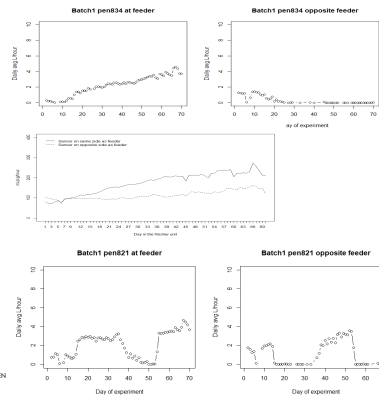
OTHER FEATURES TO EXPLORE

Two water cups:

Not always equally used

Relation to tail damage risk?

Relation to growth?



DEVELOPMENT OF WARNING ALGORITHM

Step 3: Use data to develop warning

- Use a combination of water and temperature sensors
- Next presentation by Yuvraj Dumon using machine learning
- Input data - same data set
- Specific warning of tail damage events and pen fouling

Summary and conclusion

- Behavioural changes do occur prior to pen fouling and tail damage:
- Pen fouling:
 - lying pattern- may be reflected in pen temperature
- Tail damage:
 - tail posture and activity - latter may be reflected in pattern of water consumption
- Demonstrate potentials for developing specific warning algorithms for both pen fouling and tail damage

2.2.5 Using machine learning to predict tail-biting, fouling and diarrhea in pigs

Presentation by Yuvraj Domun

Using machine learning to predict tail-biting, fouling and diarrhea
in pigs

Yuvraj Domun

Supervisory team

- Dr. Tom Norton (KU Leuven, HAU)
- Dr. Lene Juul Pedersen (Aarhus University)
- David White (HAU)
- Dr Ian Moorcroft (HAU)



12/11/2018

1

Overview

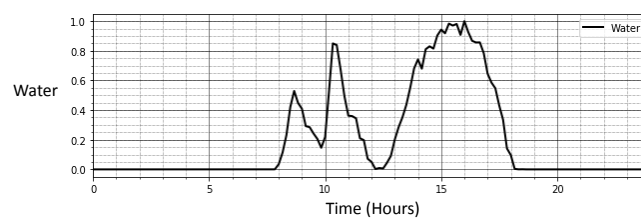
- The research problem
- The Solutions
- The Results
- Future Directions

12/11/2018

2

The Problem

Daily Water Consumption

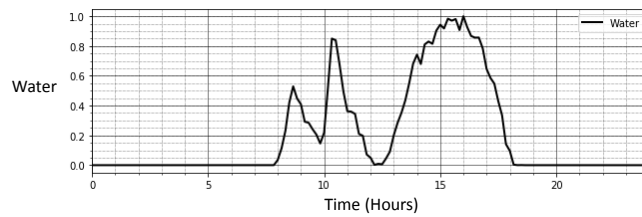


12/11/2018

3

The Problem

Daily Water Consumption



Is this a normal pattern ?

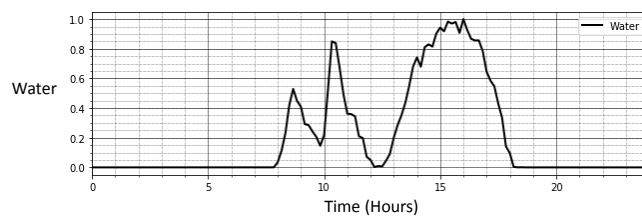
Does it indicate a specific event Y ?

12/11/2018

4

The Problem

Daily Water Consumption



Is this a normal pattern ?

Discriminate Prediction

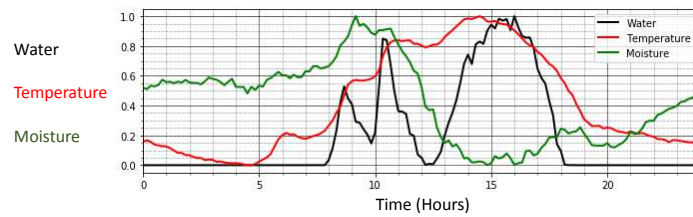
Does it indicate a specific event Y ?

12/11/2018

5

The Problem

Daily Water Consumption



Is this a normal pattern ?

Discriminate Prediction

Does it indicate a specific event Y ?

12/11/2018

6

The Problem

Develop systems that can predict specific events.

How?

- What patterns predicts certain events?

Why?

- Targeted treatment.
- Predictive Model.
- More accurate information for farmers.

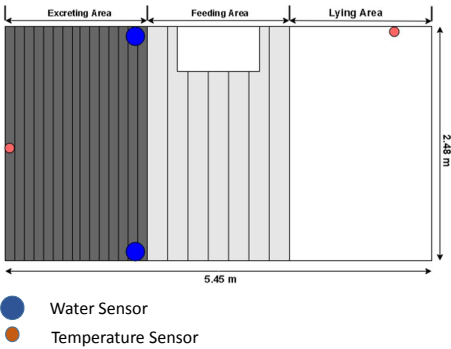
12/11/2018

7

The Problem

Data Overview (Pen layout)

Variable Name	Sampling Frequency	Unit
Temperature (Solid Floor)	1 Minute	Degrees
Temperature (Slatted Floor)	1 Minute	Degrees
Water (Drinker 1)	10 Seconds	Litres
Water (Drinker 2)	10 Seconds	Litres
Relative Humidity (Finisher Unit)	1 Minute	%
Ventilation Output (Finisher Unit)	1 Minute	%
Heating Output (Finisher Unit)	1 Minute	%
Cooling Output (Finisher Unit)	1 Minute	%
Temperature (Finisher Unit)	1 Minute	Sensor



12/11/2018

8

Solution

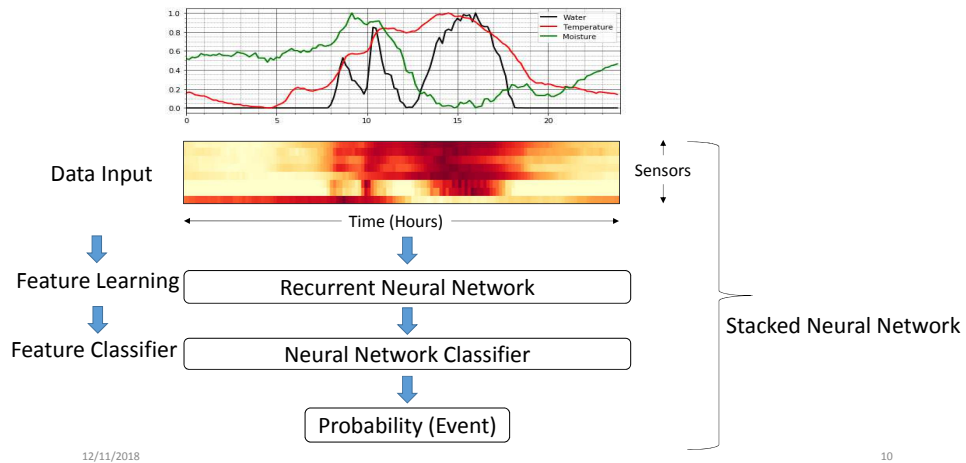
Two Proposed Solution:

1. Deep Learning Approach (Automatic Feature Learning)
2. Machine Learning Approach (Manual Feature Extraction)

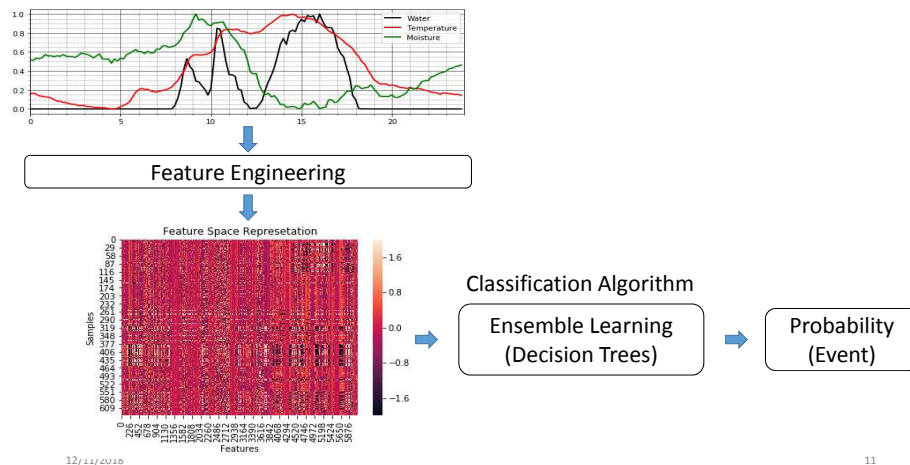
12/11/2018

9

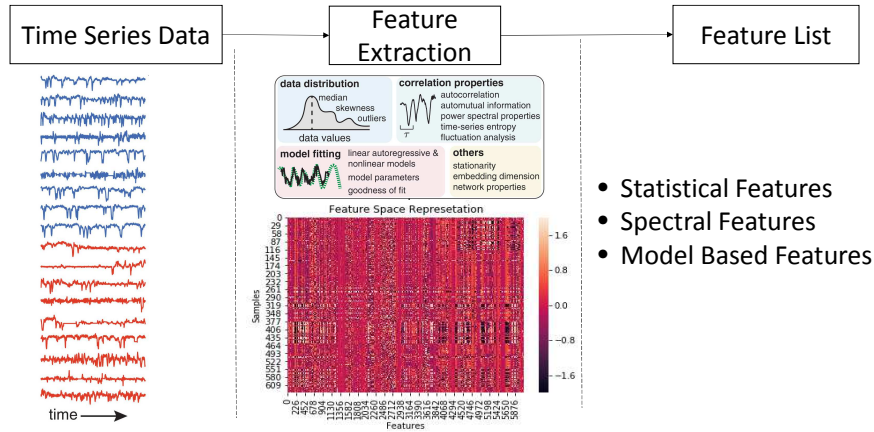
Solution 1 : Deep Learning Approach



Solution 2 : Machine Learning Approach



Solution 2 : Feature Engineering



12/11/2018

12

Results

1. Model Performance
2. Important Data Type (Sensor)
3. Important Patterns

12/11/2018

13

Results

Prediction Performance on Unseen Data

Table 1: Predictive Performance (AUROC) for fouling, tail-biting and diarrhea prediction.

Event	Deep Learning	Machine Learning
Tail Biting	0.709	0.762
Fouling	0.812	0.879
Diarrhea	0.679	0.862

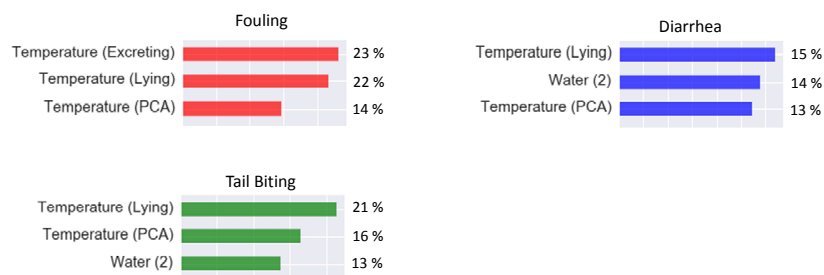
Performance Metric: Area Under the Receiver Operating Characteristic curve (AUROC)

12/11/2018

14

Results

What Sensor Data are important?



Principal Component Analysis (PCA) was applied to highly correlated features to create new features to take into account collinearity. The Feature importance (%) is the total information used by the algorithms to make the final prediction.

12/11/2018

15

Results

What Features are important?

Table 3: Top 3 most important features for predicting Fouling, Diarrhea, and tail-biting.

	Features	Feature Importance
Fouling	Fast Fourier Transform Coefficients	43%
	Quantiles Distribution	10 %
	Entropy	8 %
Diarrhea	Fast Fourier Transform Coefficients	63 %
	Quantiles	7 %
	Continuous Wavelet Transform Coefficients)	7 %
Tail Biting	Fast Fourier Transform Coefficients	54 %
	Linear Trend	11 %
	Entropy	5 %

Principal Component Analysis (PCA) was applied to highly correlated features to create new features to take into account collinearity. The Feature importance (%) is the total information used by the algorithms to make the final prediction.

12/11/2018

16

Conclusion

- Frequency based Patterns are important.
- Traditional machine learning approach better than Deep Learning.
- Both approaches can learn to predict discriminate events.

Future Studies

- Integrate proposed approach with more advance sensors.
- Can be extended to detect other events beside fouling, diarrhea, or tailbiting.

12/11/2018

17

Acknowledgment

This research was supported by Chadacre Agricultural Trust, John Oldacre Foundation, and the Douglas Bamford Trust.

We would equally like to thank the Danish Council for Strategic Research (The PigIT Project, Grant number 11- 116191) and the Green Development and Demonstration Programme under the Ministry of Food, Agriculture and Fisheries, Denmark (project IntactTails j.nr. 34009-13-0743) for supplying the data for this research

12/11/2018

18

Questions?

12/11/2018

19



PigIT Report No. 10 • November 2018
<http://www.pigit.net/publications/PigIT-Report10.pdf>

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