

Detection of Dairy Herds at risk for changing Salmonella Dublin status

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Publication date: 2012

Document Version Publisher's PDF, also known as Version of record

Link back to DTU Orbit

Citation (APA): Stockmarr, A. (Author), Bødker, R. (Author), & Nielsen, L. (Author). (2012). Detection of Dairy Herds at risk for changing Salmonella Dublin status. Sound/Visual production (digital) http://www.statistiksymposium.dk/

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Detection of Dairy Herds at risk for changing *Salmonella* Dublin status



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Symposium in Applied Statistics 2012

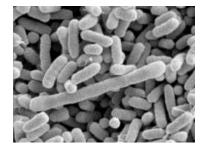
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DTU Informatics Department of Informatics and Mathematical Modeling

Background

Salmonella Dublin (S. Dublin) is a specific serotype of the Salmonella bacterium



- S. Dublin is host-adapted to cattle, and the most prevalent serotype found in cattle in Denmark (~60-70% of all isolates)
- S. Dublin is a rare but serious zoonosis that causes severe disease and deaths in humans every year (10-40 hospitalized cases per year in Denmark).

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S. Dublin Symptoms

- Unthrifty calves Fever Diarrhoea (bloody)
 - Pneumonia Death Abortions







The Salmonella Dublin surveillance and eradication programme in Denmark



- Eradication campaign until 2014
- Goal: The Danish cattle population free from Salmonella Dublin in 2014



- 2010-2012: Sanctions to improve motivation
- 2013-2014: Veterinary Authorities will handle infected herds through law enforcement

The Salmonella Dublin surveillance and eradication programme in Denmark

- Surveillance program since 2002:
- Cattle herds are classified as follows, based on Bulk Tank Milk/blood sample antibodies and trade contacts:
- Level 1: Most likely **free** from salmonella
- Level 2: **Too high antibody** levels or contact to other herds in Level 2 or 3
- Level 3: **Clinical** *Salmonella* Dublin diagnosis and culture positive
- Unknown: Only non-dairy herds with too **few samples** to classify (hobby herds)

The Salmonella Dublin surveillance and eradication programme in Denmark

- Surveillance program since 2002:
- Cattle herds are classified as follows, based on Bulk Tank Milk/Blood sample antibodies and trade contacts:

Level 1: Most likely free from salmonellaLevel 2: Too high antibody levels or contact to other herds in Level 2

For the present work we do not distinguish between level 2 and 3

Reasons for assigning Level 1 to dairy

Salmonella Dublin Level 1 is given if:

- Valid bulk tank milk antibody measurements exists

 and
- The last 4 bulk tank milk antibody measurements, gathered with at least 3 weeks in between, shows an average ODC-value of less than 25 – and
- 3. The latest Salmonella Dublin measurement has **not shown an increase of more than 20**, compared to the average of the three preceding measurements - and
- 4. A number of circumstances mainly related to trade and missing data do not hold.

Otherwise, Salmonella Dublin Level 2 is given.



Bulk Tank Milk antibody measurements from dairy herds



- Collected routinely every three months.
- Thus, a very long period for an infection to develop before a dairy herd is possibly re-classified.
- Sanctions and law enforcement gives farmers an incentive to act if they suspect an infection is present.
- There is therefore a need to identify herds at risk of changing S. Dublin level, based on information a quarter earlier than reclassification takes place.



Risk Herds

- Are herds 'at risk of' changing level from 1 to 2; we will start out with this loose definition.
- Risk herds are thus Level 1 ("Status 1") herds.

Purpose of current study:

• To determine appropriate definitions for a risk herd, based on available factors one quarter prior to a possible level shift.

The probability of changing *S.* Dublin status

If $X_{i,t}$ denotes the BTM measurement for herd *i* at time *t*, and $M_{i,t}$ the mean of the last 3 measurements, $M_{i,t} = \frac{1}{3} \sum_{j=1}^{3} X_{i,t-j}$ then the probability $P_{i,t}$ of a status change is given as $p_{i,t} = P(\frac{1}{4}X_{i,t} + \frac{3}{4}(\frac{1}{3}\sum_{i=1}^{3}X_{i,t-i}) > 25 \lor X_{i,t} - (\frac{1}{3}\sum_{j=1}^{3}X_{i,t-j}) > 20 | X_{i,t-1}, X_{i,t-2}, X_{i,t-3})$

ie.

$$p_{i,t} = P(\frac{1}{4}X_{i,t} + \frac{3}{4}M_{i,t} > 25 \lor X_{i,t} - M_{i,t} > 20 | M_{i,t})$$

 $p_{i,t}$ is modeled through a logistic regression: $\log it(p_{i,t}) = \sum \beta_k Z_{i,k,t}$

through successive conditioning on the BTM measurements.

What makes herds become Risk Herds? - bulk tank milk measurements, trade, neighbors and herd size

- Bulk tank milk measurements:
 - -High antibody levels;
 - -Unstable development in measurements.
- Trade:
 - That animals are bought from herds that turn out to be Status 2 herds;
 - -that many animals are bought;
 - -that many herds are traded with.
- Neighbors:
 - -That many neighbors are assigned Status 2;
 - -That there are many neighbors.
- Size:
 - -That the herd is large.



Available Data

- Quarterly bulk tank milk measurements 2002-2008 for 9397 dairy herds;
- Geographical coordinates for dairy and non-dairy cattle herds and their quarterly *S*. Dublin level;
- All perfomed trades at animal level 2002-2008;
- Data on herd sizes; Only 2004-2008 er usable.



BTM measurements and Alarm Herds

- BTM measurement enter covariates through $M_{i,i}$; obviously a high level of recent antibody levels will increase the risk of a Status 2 change.
- But also sudden (upwards) deviations from a stationary development could indicate an emerging infection. BTM measurements also enters through *Alarm Herd* status:
- Alarm Herd concept: A Status 1 herd A is an <u>Alarm Herd</u> at a timepoint t_0 , if BTM measurements for A for at least the previous 4 time points do not vary more than a standard 95% confidence interval would predict, and the BTM measurement for A at time t_0 is above this level, and above an upward threshold c > 0.

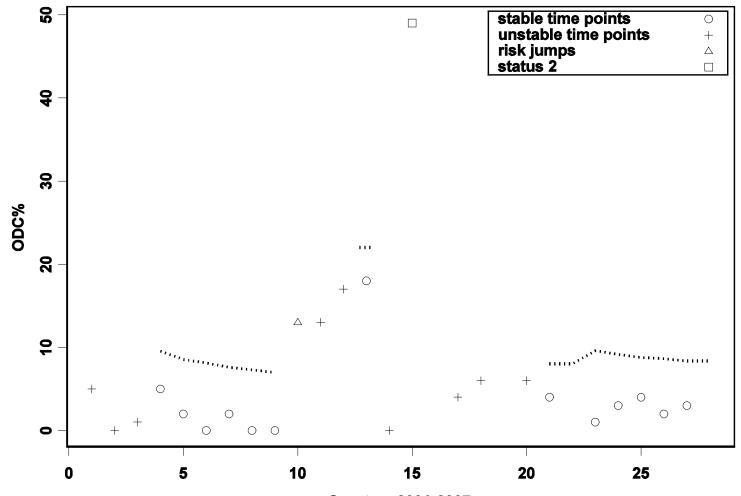


Alarm Herds

- Alarm Herds are Level 1 herds: The jump from the steady progression should be big (to a level > c), but not so big that it triggers a Status 2 classification.
- Time points where BTM measurements for A for at least the previous 4 time points do not vary more than a standard 95% confidence interval would predict are called <u>stable timepoints</u>.
- When a herd leaves a stable state to become unstable, we say that a <u>jump</u> has occurred.
- When a herd becomes an Alarm Herd, we say that a <u>risk</u> jump has occurred.



A BTM measurement progression example



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Estimation of c



- We included Alarm Herd status in the basic logistic regression model with varying values of c (integer ODC values), to gain a series of competing models;
- We chose the model with the optimal Akaike Information;
- Consequently, we estimated *c* to be 13.

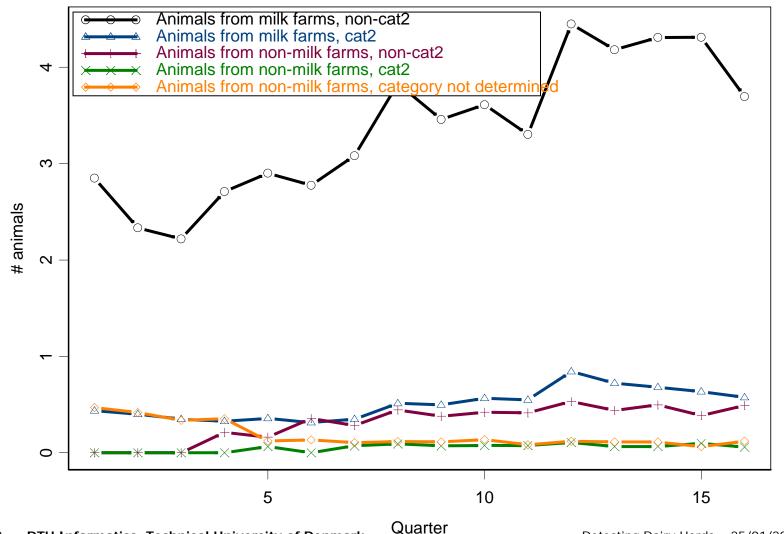
Trade information







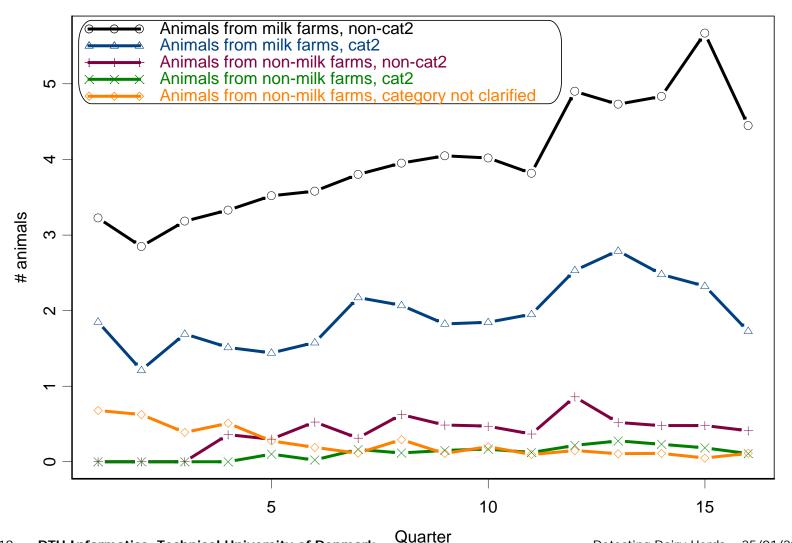
Milk producing non-cat 2 herds purchases 2004-2007



Trade II



Milk producing cat 2 herds purchases 2004-2007



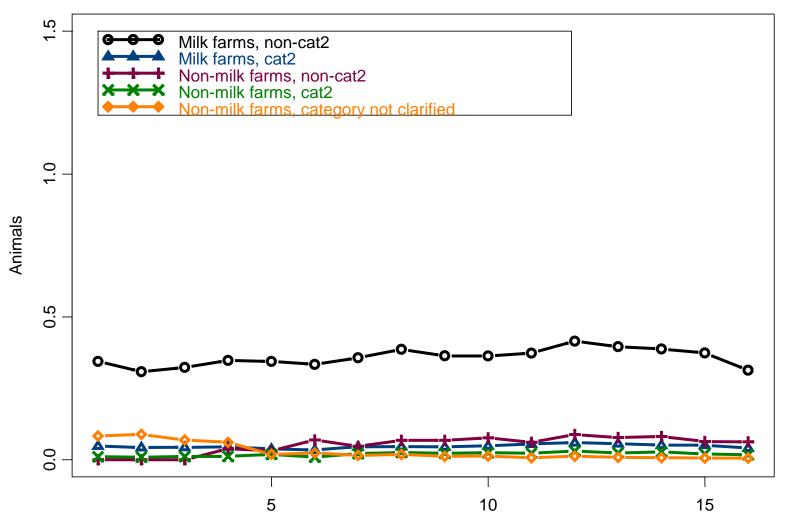
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Trade III

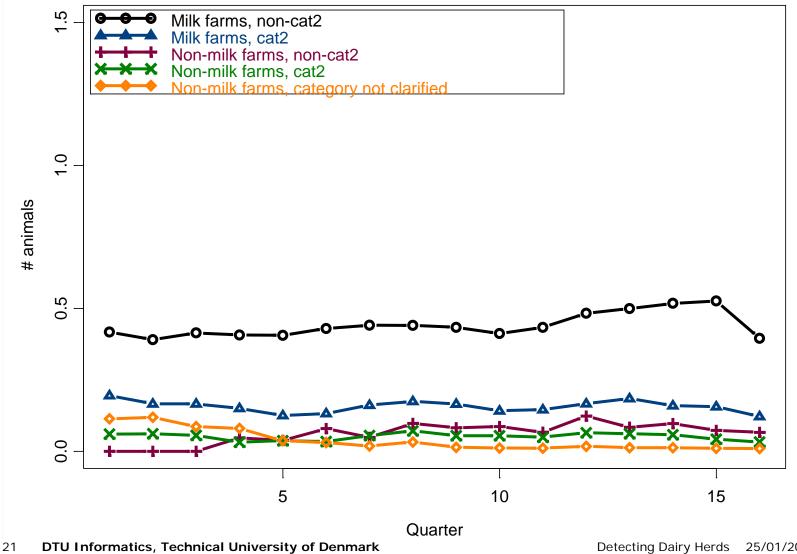


Trade partners for milk producing non-cat 2 farms 2004-2007



Trade IV

Trade partners for milk producing cat2 farms 2004-2007



Neighbor herds and animals





Neighbors – What is a neighbor?

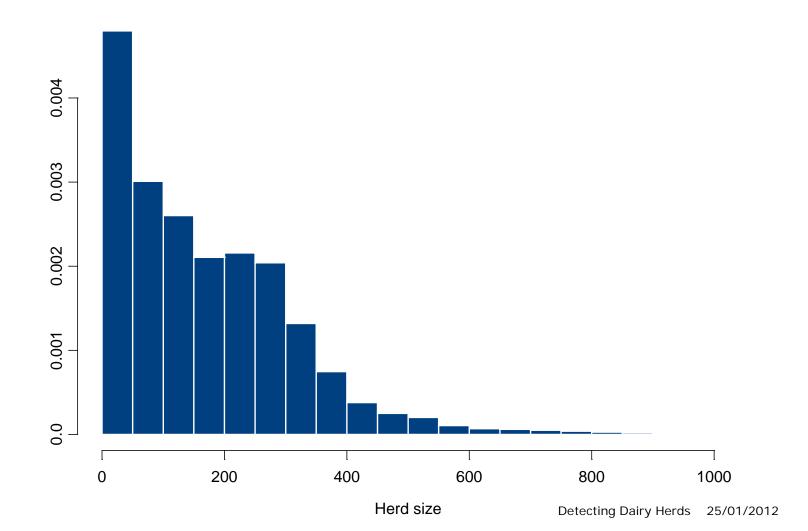


- Ersbøll & Nielsen (2008): 4.9 km is the average 'range of influence' for local spread of Salmonella.
- We define <u>Neighbors</u> to be herds with a distance of less than 4.9 km from the herd in question.
- Counted for all dairy herds in Denmark, dynamically (ie., a time series).
- We consider both the number of herds and the number of animals within this radius.
- The neighbor effect is a hidden geographical component, in that clusters of herds will have many Neighbors (well known places in Jutland).
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Herd Sizes, dairy herds

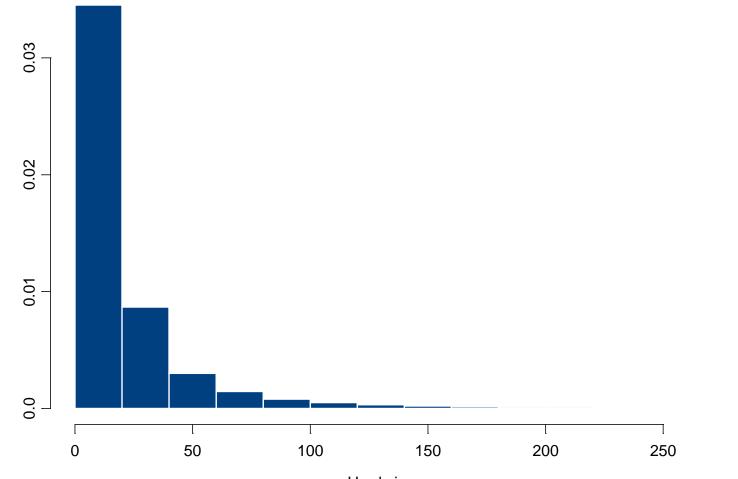


Histogram, milk producing herds



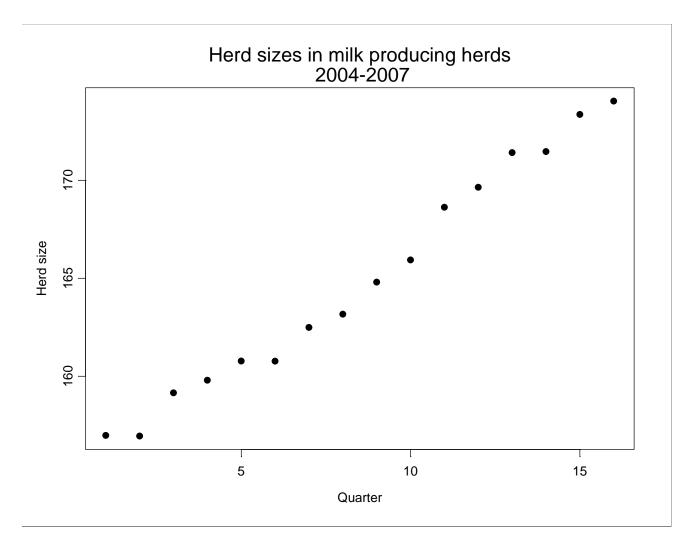
Herd sizes, non-dairy herds (used as neighbors only)

Histogram, milk producing herds



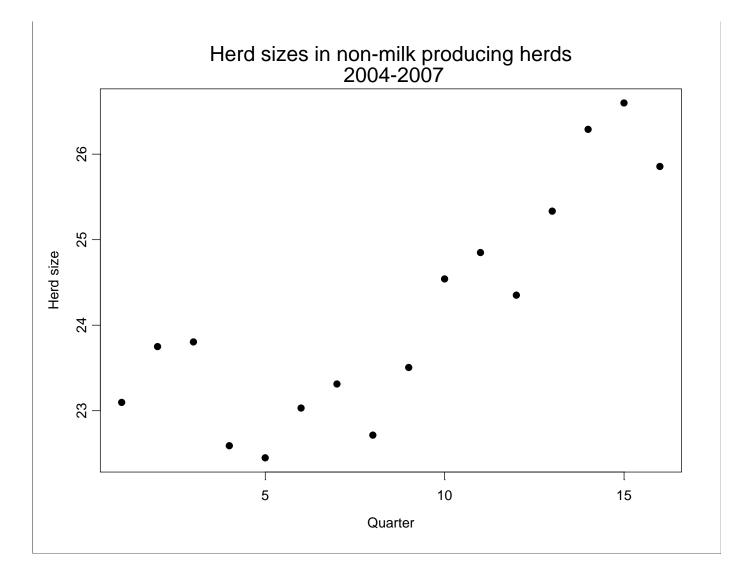
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Dynamic herd sizes, dairy herds



Dynamic herd sizes, non-dairy herds





Explanatory variable	Type of effect	Mean of explanatory variable	Regression coefficient ± S.E.	Р
Intercept	Main effect	1	4.90 ± 0.20	<0.0001
Season, ^{1st} quarter	Seasonal effect	0.27	-6.80 ± 0.30	<0.0001
Season , 2 nd quarter	Seasonal effect	0.26	-7.14 ± 0.32	<0.0001
Season, 3 rd quarter	Seasonal effect	0.26	-6.62 ± 0.29	<0.0001
Mean of last 3 (Mean3)	Main effect	5.34	0.12 ±0.016	<0.0001
Mean3, 1 st quarter**	Interaction	5.25*	0.071± 0.021	<0.0001
Mean3, 2 nd quarter	Interaction	5.27*	0.11 ± 0.022	<0.0001
Mean 3, 3 rd quarter	Interaction	5.47*	0.067 ± 0.021	<0.0001
Trade contacts with Status 2 herds	Main effect	0.037	0.44 ± 0.21	-
Animals traded with Status 1 herds	Main effect	2.62	$5.30e-3 \pm 3.44e-3$	-
Animals traded with Status 2 herds	Main effect	0.29	0.033 ± 0.010	-
Neighbour Status 2 dairy herds	Main effect	4.28	0.066± 0.033	-
Neighbour animals from Status 2 dairy herds	Main effect	816	$4.01e-4 \pm 1.61e-4$	-
Alarm1	Main effect	0.019	0.58± 0.32	0.001
Mean3 × Animals traded with Status 2 herds	Interaction	2.51	$-1.14e-3 \pm 6.41e-4$	<0.0001
Mean 3 × Neighbour animals from Status 2 dairy herds	Interaction	5370	$-1.27e-5 \pm 5.70e-6$	<0.0001
Trade contacts with Status 2 herds × Animals traded with Status 2 herds	Interaction	0.40	-9.33 e-3± 6.21e-3	<0.0001
Neighbour Status 2 dairy herds × Neighbour animals from Status 2 dairy herds	Interaction	8250	-1.69e-5 ± 0.73e-6 Detecting D	Dairy Herds 25/01/201

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Remarks I



- No effect of trade two quarters back.
- Interaction across groups of covariates with M_{t} only.
- Alarm Herd status (one quarter back) has a large impact;
- Effect of animals (traded and neighbors) decreases when level of M_{τ} increases; those with a high level of M_{τ} have a risk independent of animals purchased and animals traded.
- Independent of M_{t} :

Trade contacts with Status 2 herds and Alarm Herd status.

Remarks II



- Trade with Status 1 herds has an effect through the number of purchased animals, while trade with Status 2 herds has an effect both through the number of trade partners and the number of purchased animals.
- Status 1 trade contacts is not significant.
- Neighbors has an effect through Status 2 dairy neighbor farms and Status 2 dairy animals.
- Non-dairy neighbours and Status 1 dairy neighbors are not significant.

Model based on 2007 data only

Table 2 Explanatory variables, regression coefficients and P-values in the final logistic regression model

for a change in herd classification from Status 1 to Status 2 in the Danish surveillance programme for S.

Dublin in dairy herds, base on data from 2007

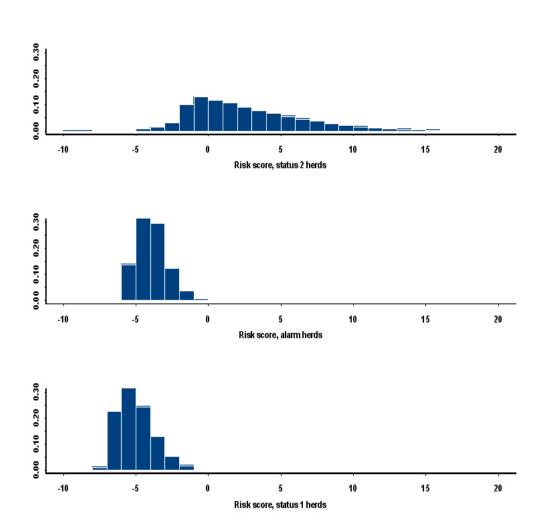
Explanatory variable	Type of effect	Mean of explanatory variable	Regression coefficient ± S.E.	р
Intercept	Main effect	1	-7.05 ± 0.39	<0.0001
Mean3	Main effect	6.50	0.22 ± 0.02	<0.0001
Trade contacts with Status 2 herds	Main effect	0.048	0.70 ± 0.37	0.0006
Number of animals in neighbouring Status 2 dairy herds	Main effect	758	0.0179 ± 0.0113	0.003
Alarm1	Main effect	0.023	1.08 ±0.69	0.006

Risk Scores

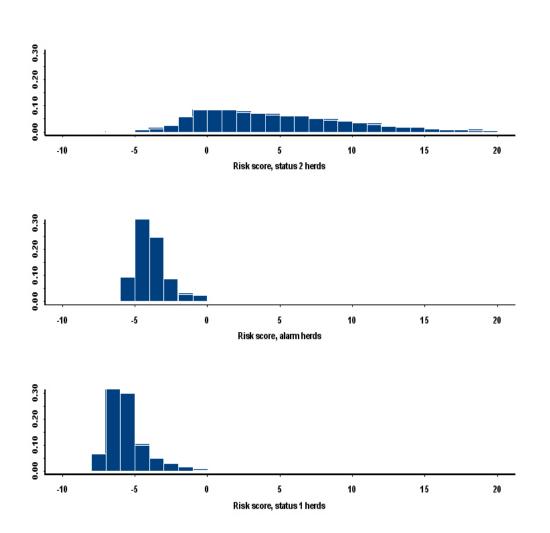


- We used the linear predictor as a risk score index.
- Which index value is high enough to consider a given herd to be a 'risk herd', where the farmer should intervene if possible?

Risk Scores – big model



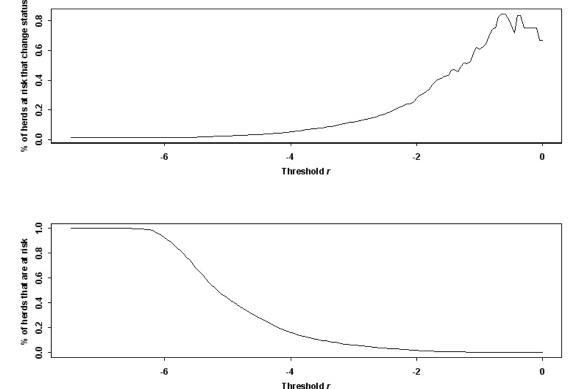
Risk Scores – small model





Definitions of risk herds

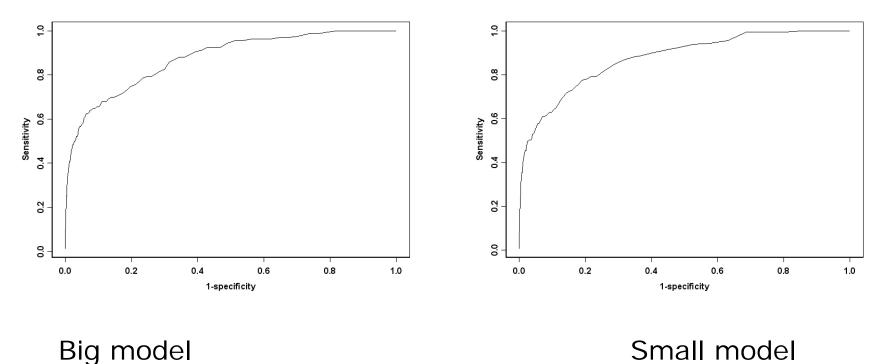
 A threshold *r* that defines a herd to be a risk herd if the risk score is > *r* represents a trade-off between number of herds at risk and the frequency with which they change status:



Predictive power



Coefficients re-estimated based on 2004-2006 data, and used to predict 2007 data: ROC curves when threshold varies





Small model versus big model

- Tempting to choose the simpler model;
- Predictive power is similar and model is much less complicated;
- However, the model could not be reproduced in full when applied to different cohorts of the data; we have no explanation for this.
- We recommend use of the big model.

Predictive power and relative importance



 Threshold is optimized based on predictive power of the big model as

$$r = -1.05$$

- However, it is much more important to predict those that change status than those that do not.
- Let α denote the importance of predicting a status change correctly relative to a non-change, and let C be the event "change of status", and PC the event that a herd is classified as a Risk Herd. Instead of optimizing average predictability, we optimise the importance function

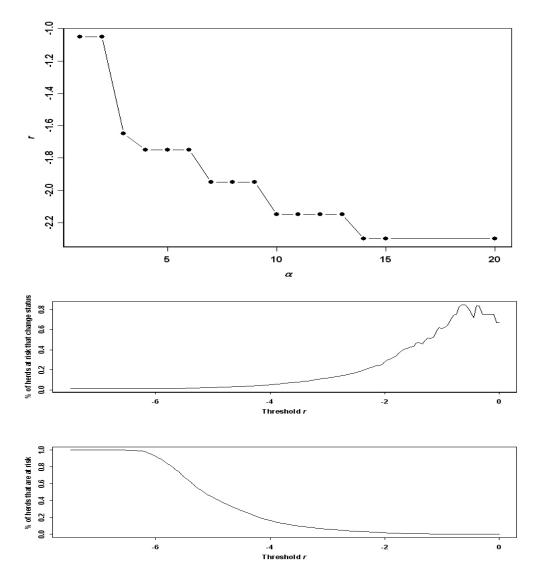
$\alpha P(C|PC)P(PC)+P(C^{\neg}|PC^{\neg})P(PC^{\neg})$

where " ¬ " signifies negation.

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Threshold value as a function of importance





Suggested estimation of α based on farmer incentive

• Loss if a Status 2 herd is not correctly predicted:

Price[3 months in Status 2]

-Price[3 months in Status 2]*

P(Interventions fail)

- -Price[Interventions]
- Loss if a Status 1 herd is not correctly predicted:

Price[Interventions]

$\hat{\alpha} = \frac{\text{Price}[3 \text{ months in Status 2}]*P(\text{Interventions succeed})}{\text{Price}[\text{Interventions}]} - 1$

Suggested estimation of α based on societal costs

Replace Price[3 months in Status 2] by

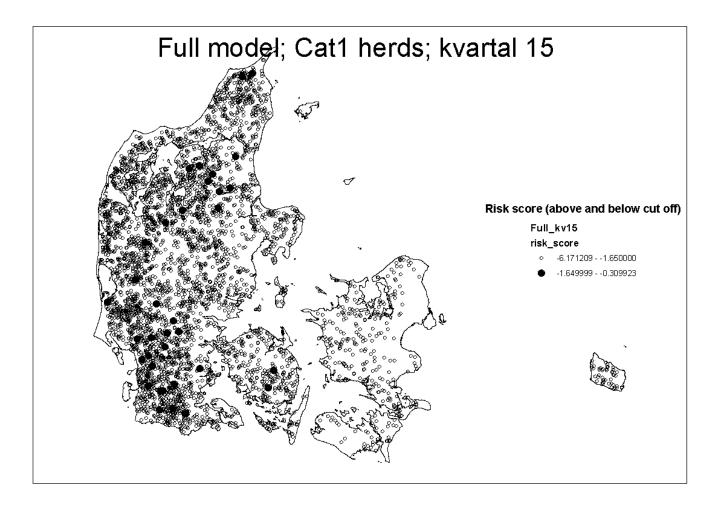
R₀[Price[3 months in Status 2]
+cost(human infections) per herd]

where R_0 denotes the average excess number of infected herds due to delayed identification.

Higher α and lower threshold *r*.

These cost values are not known and dependent on legislation

Geographical distribution of risk herds ($\alpha = 5$) $\stackrel{110}{\rightleftharpoons}$ 3rd quarter 2007.



Conclusion



- We suggest that risk herds may be defined as herds with a risk score over a threshold *r*, with *r* taking a value from -1.05 and lower, depending on the nature of sanctions and the importance of detecting status changes.
- Potential uses if risk index: replacement of the current classification system; potential legal conflicts should be clarified.
- Alternatively, a mandatory notice to the farmer on a risk herd classification, allowing voluntary interventions;
- Coming legislation should encourage farmers to intervene. However, cost to society is higher than the cost to individual farmers due to spread of disease, so a higher α value and thus lower threshold r could apply.



Acknowledgement

• We thank Jørgen Nielsen from the Danish Cattle Federation for providing register data from the Danish Cattle Database. Thanks Jørgen.