



Faculty of Health and Medical Sciences



Dynamic monitoring of mortality rate for sows and piglets

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Dias 1





Background

Problem

- Existing Management Information Systems (MIS) are *static* and typically computed every quarter or year

Solution

- Dynamic approach

- Save time
- Save money



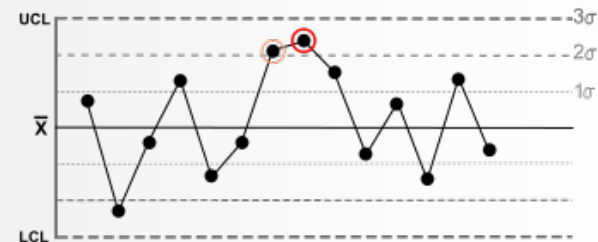


Objective



Develop new and more reliable methods for dynamic production monitoring

- Implementation of automatic methods for detection of systematic deviations from the expected results





Mortality rate model

The third phase of the project:

- Mortality rate being a binary trait, is modeled by using a Dynamic Generalized Linear Model (**DGLM**)
- Mortality rate of sows and piglets is treated at the same time

Model and dataset components:

- 15 parameters $\rightarrow \mu_t, \alpha_{2t}$ to $\alpha_{8t}, \beta_{2t}, \gamma_{1t}$ to $\gamma_{4t}, \zeta_t, \delta_t$
- 15 herds
- 3-9 years





Model parameters

μ_t → general level for sow mortality

α_{2t} to α_{8t} → coeff. referring to parity effects on sow mortality

β_{2t} → effect of the stage in the reproductive cycle
(*insemination + gestation and nursing + dry*)

γ_{1t} to γ_{4t} → coeff. for the parity specific stillbirth rate

ζ_t → coeff. for pre-weaning mortality

δ_t → coeff. for the slope of stillbirth rate after parity 4

The parameter vector will be:

$$\theta_t = (\mu_t, \alpha_{2t}, \dots, \alpha_{Nt}, \beta_{2t}, \gamma_{1t}, \dots, \gamma_{4t}, \zeta_t, \delta_t)'$$



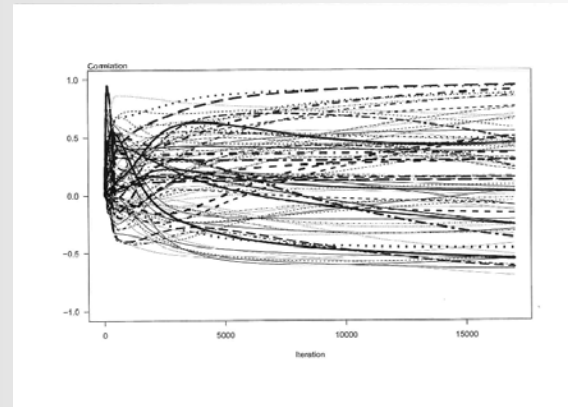
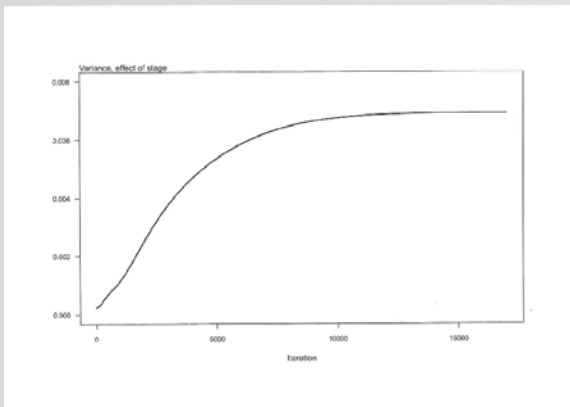
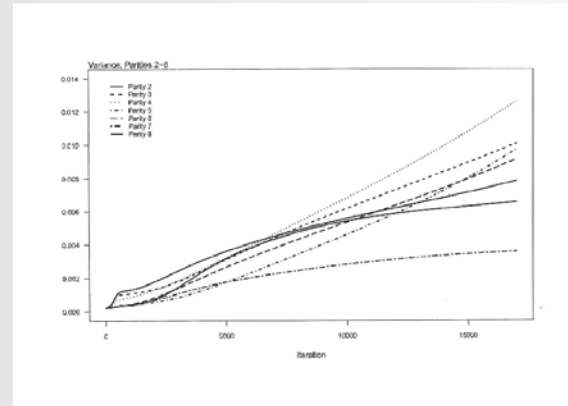
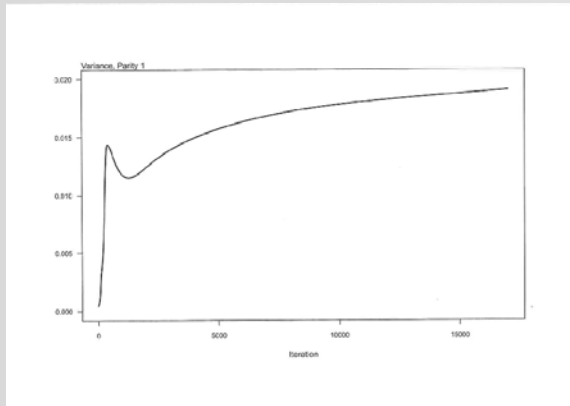


Material and methods

- Once that the parameters (μ_t , α_{2t} to α_{8t} , β_{2t} , γ_{1t} to γ_{4t} , ζ_t , δ_t) were calculated, at any week the last week's estimate is available as a result of the **Kalman filter** application
- Expectation-Maximization (**EM**) **algorithm** technique was used for estimation of the system variance
- Detection methods (**Control chart** and **V-mask**) were applied in order to monitor out-of-control situations



Results – Variance & Correlation sows - EM



???

Correlation
between all
parameters

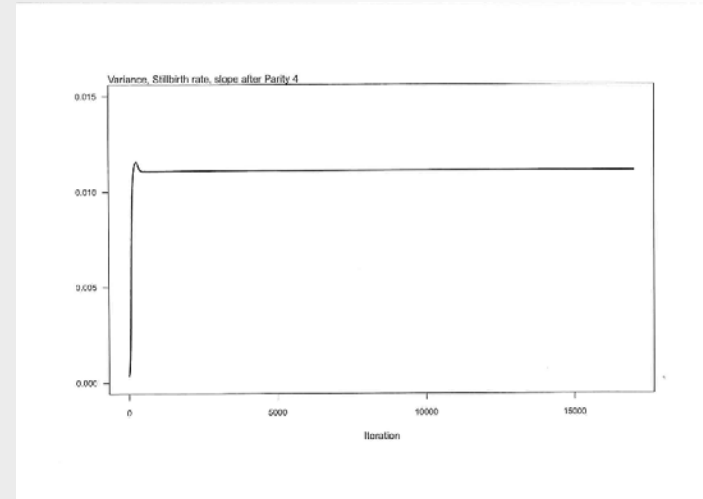
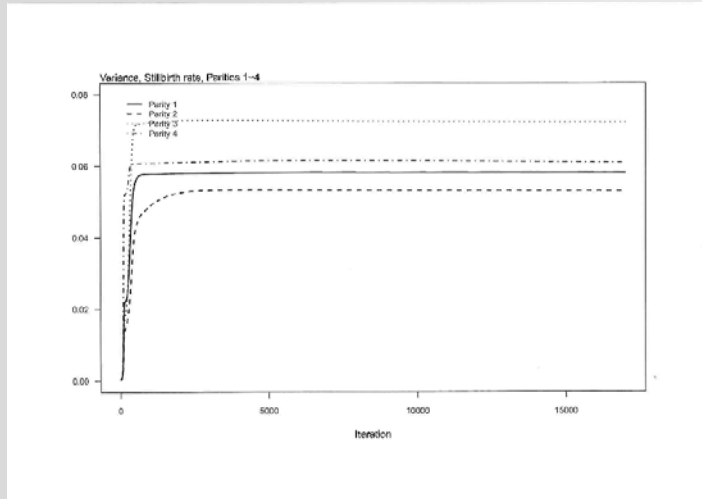
- Convergence: 17000 iterations
- Variance
- Correlation between parameters (...)

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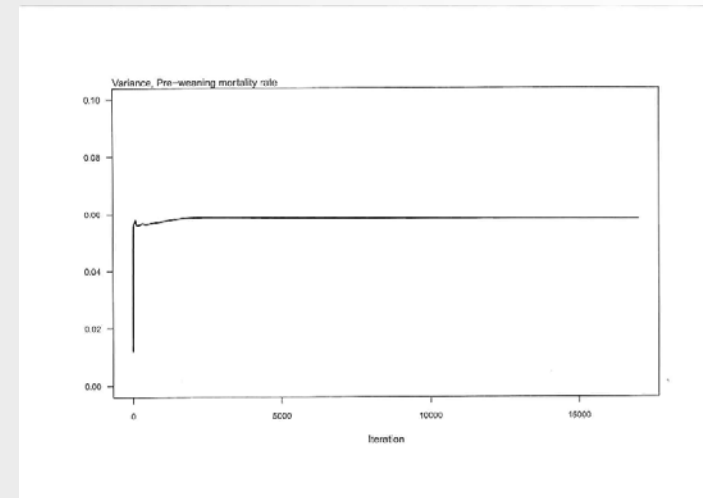


Results – Variance & convergence piglets - EM



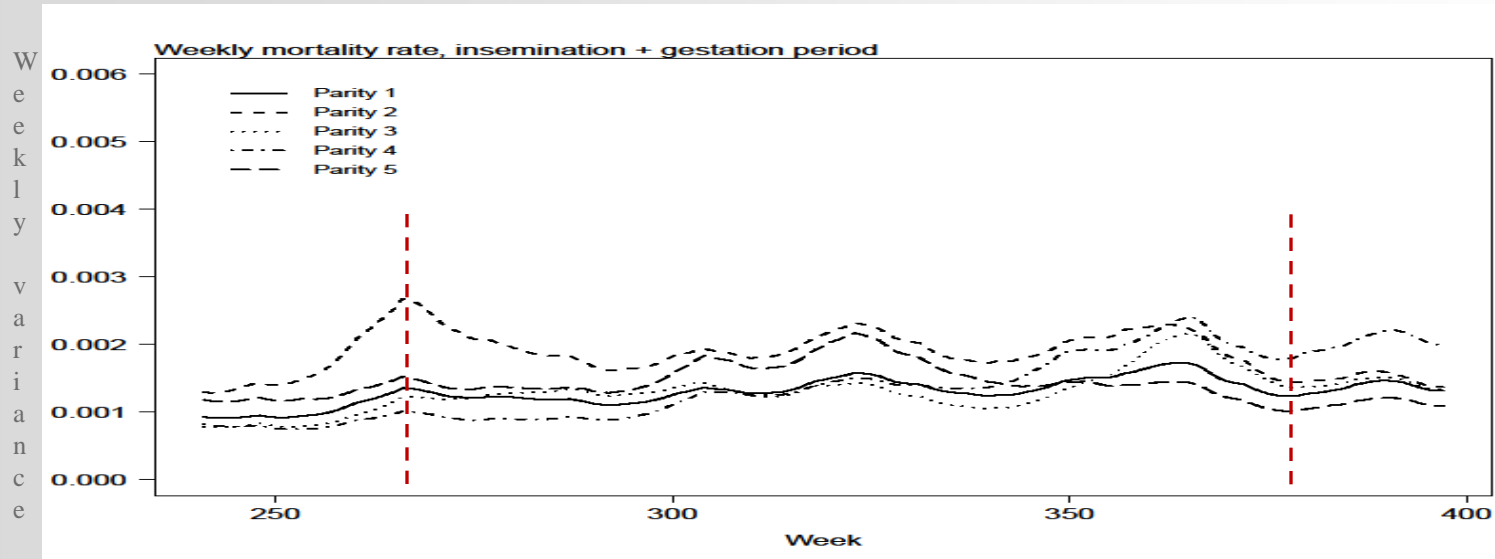
Convergence has been reached after less than 5000 iterations for:

- Stillbirth rate 1-4 parities
- Stillbirth slope after parity 4
- Pre-weaning mortality





Results – Mortality rate per parity (3 years) – sows KF



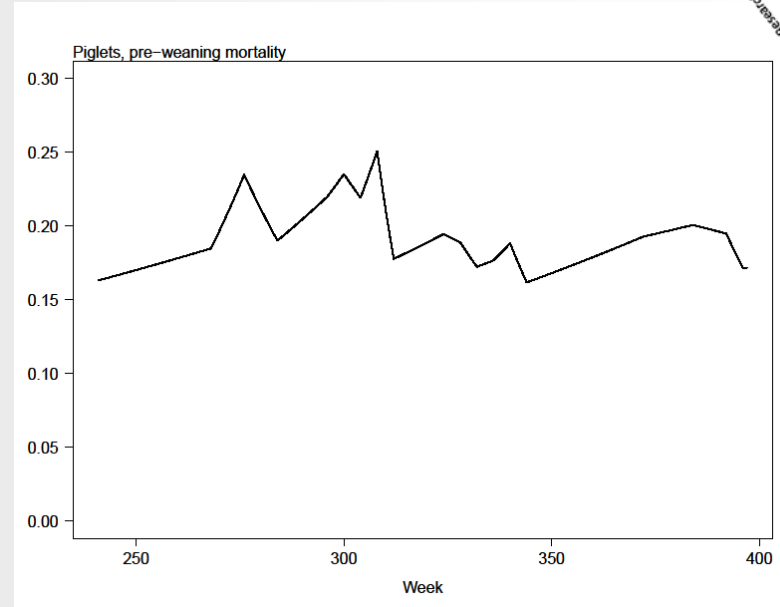
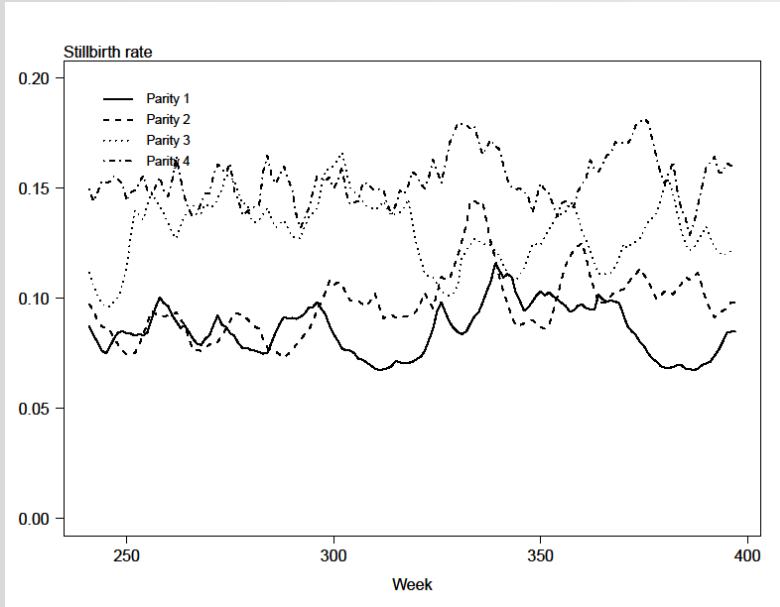
Sow mortality in 3 years for the first 5 parities – smoothed data

- The smoothing represents the best possible estimate
- It is possible to recognize a pattern (correlation)
- Red dotted lines → alarms in V-mask + Cusum [later commented]





Results – Mortality rate per parity (3 years) – piglets KL



Piglets mortality in 3 years – smoothed data

- No clear patterns have been found in stillbirth rate over the first four parities
- Pre-weaning mortality ranges between 16% and 25 %



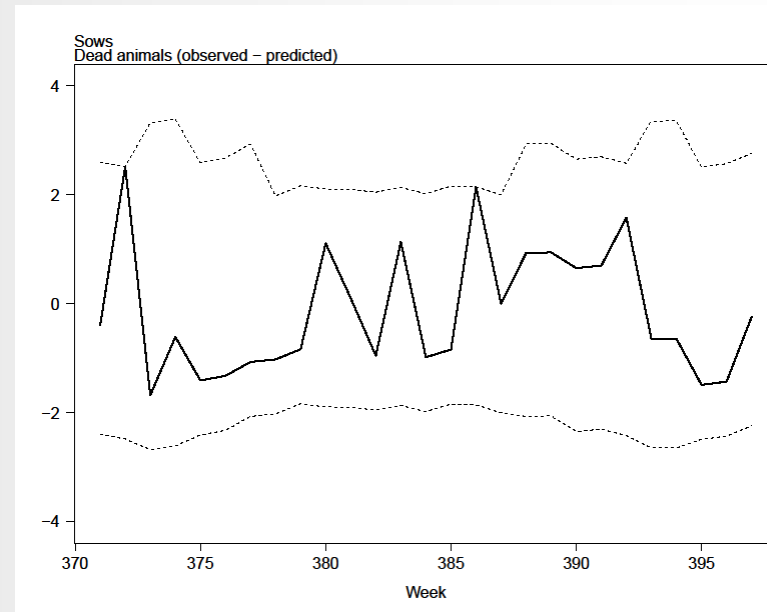
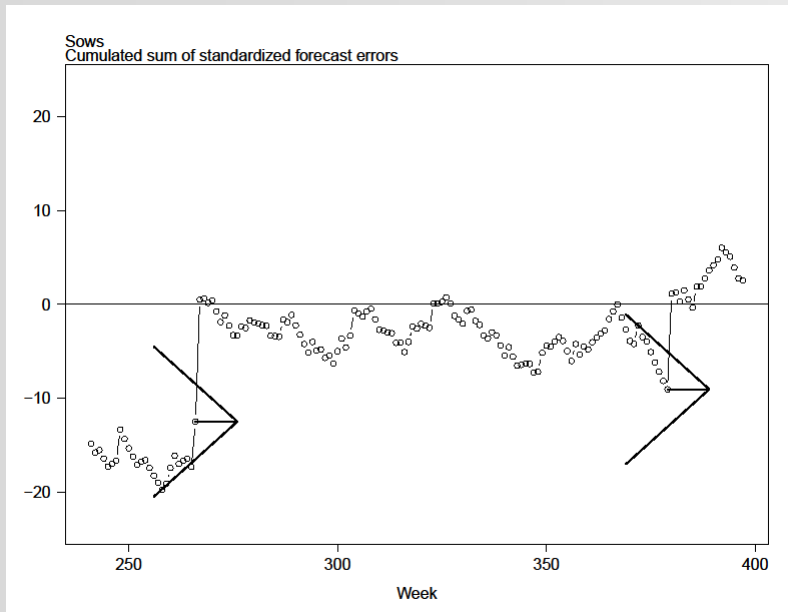


Results – Detection methods sows

Results were monitored both in a short and long time period

Monitoring methods:

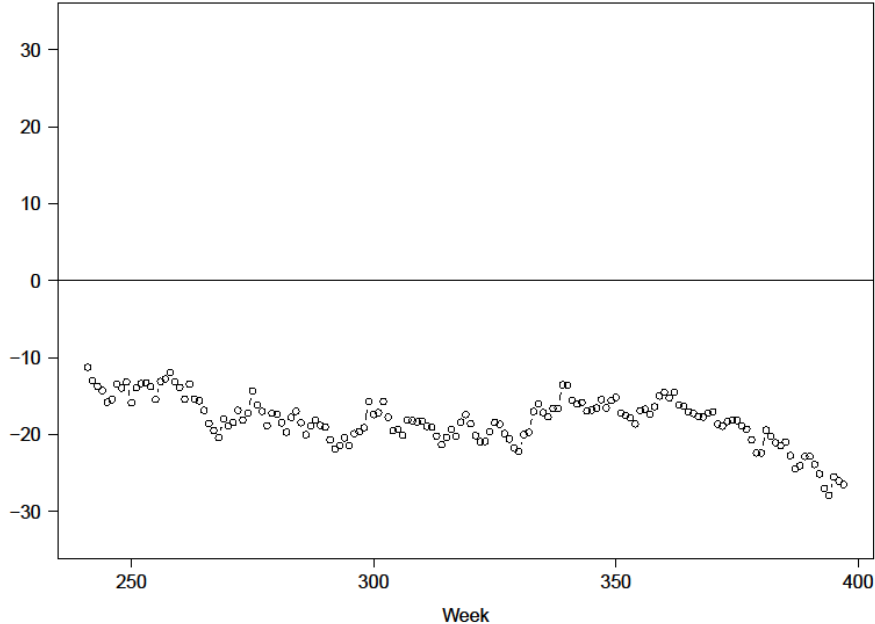
- V-mask applied on Cusum \rightarrow Long time horizon (3 years)
- Control Chart \rightarrow Short time horizon (26 weeks)



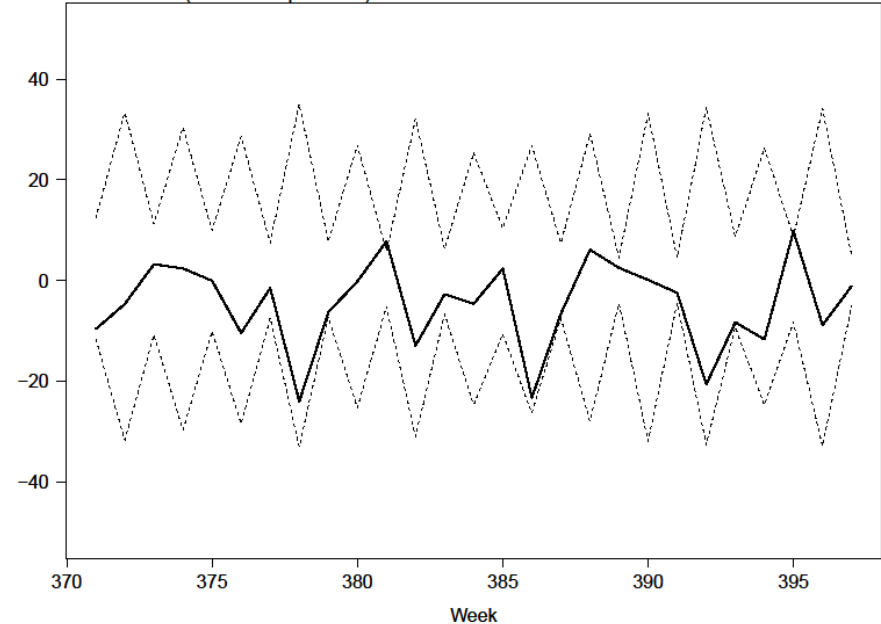


Results – Detection methods piglets

Stillborn
Cumulated sum of standardized forecast errors



Stillborn
Dead animals (observed – predicted)



No alarms have been found

→ Warning limits suggested for control chart method





Conclusions

- The system for monitoring mortality rate of sows and piglets based on DGLM, V-mask and Control Chart can be a useful tool for modeling mortality rate week by week in short and long period
- V-mask settings (angle of the arms) need to be optimized
- Parity specific deviations can be monitored
- A suggestion for further developments can be to “*split*” the model into sow mortality and piglet mortality, and combine them at a later stage



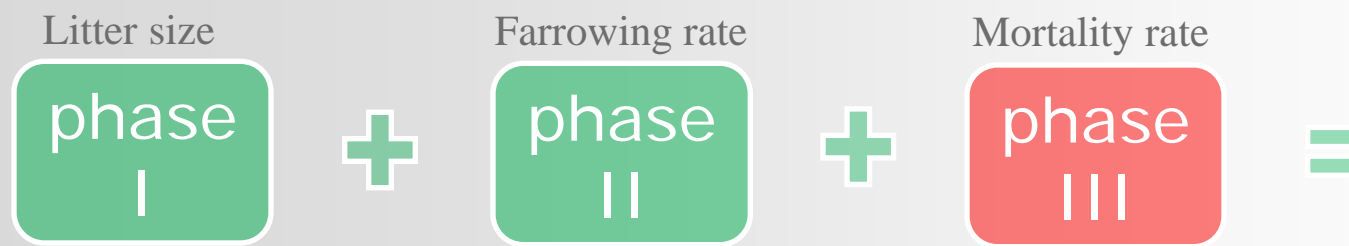


Conclusions and Perspectives

The junction of the three models will represent an important step towards the construction of a management tool (i.e. software)

This tool may be used:

- To help in the decision support system
- To monitor changes
- To predict production in a dynamic way



Thank
you





First two phases

Livestock Science 149 (2012) 289–300

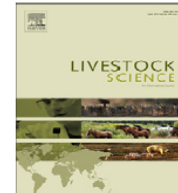


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Dynamic production monitoring in pig herds I: Modeling and monitoring litter size at herd and sow level

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Dynamic production monitoring in pig herds II. Modeling and monitoring farrowing rate at herd level





Material and methods

Dynamic Generalized Linear Model (DGLM)

DGLM consists of these equations:

Observation equations $\rightarrow y_{Gt} | \theta_t \sim \mathcal{B}(N_{Gt}, p_{Gt})$

$$\eta_t = F_t' \theta_t$$

System equation $\rightarrow \theta_t = G_t \theta_{t-1} + w_t$

\mathcal{B} \rightarrow denotes the binomial distribution

The parameter vector for week t will be:

$$\theta_t = (\mu_t, \alpha_{2t}, \dots, \alpha_{Nt}, \beta_{2t}, \gamma_{1t}, \dots, \gamma_{4t}, \zeta_t, \delta_t)'$$

