



DEPARTMENT OF ECONOMICS AND BUSINESS
AARHUS UNIVERSITY

A hierarchical Markov decision process modelling feeding and marketing decisions of growing pigs

Reza Pourmoayed and Lars Relund Nielsen

Department of Economics and Business, Aarhus University, Denmark

Anders Ringgaard Kristensen

Department of Large Animal Sciences, University of Copenhagen, Denmark

IFORS conference - July 17 , 2014



Agenda

PigIT research group

Optimal feeding and marketing decisions

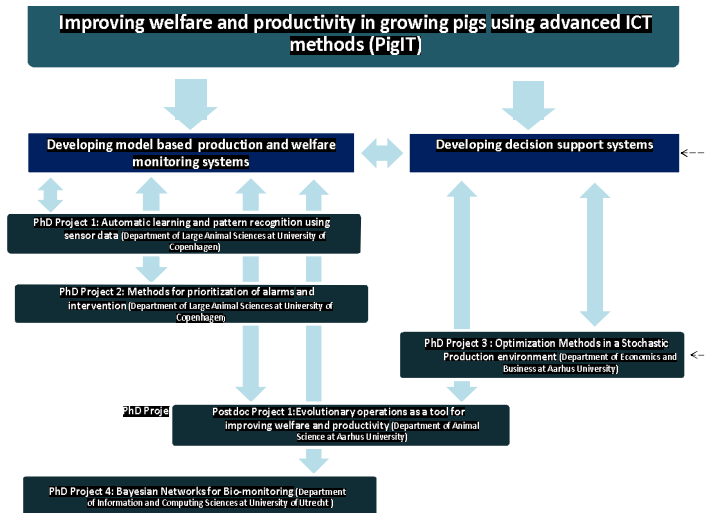
Problem definition

Modelling approach

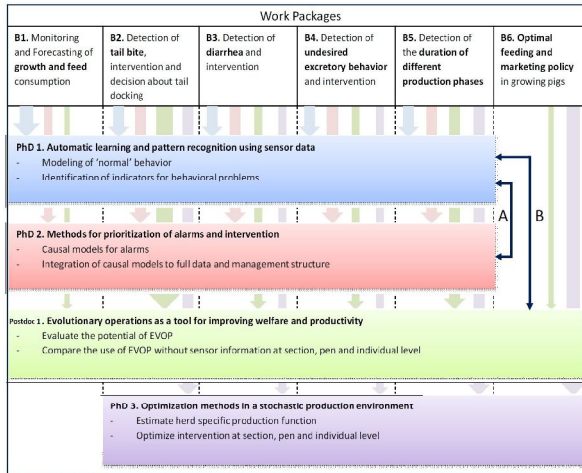
Initial results

Further research

PigIT project



Work packages



Source: <http://pigit.ku.dk/>

Agenda

PigIT research group

Optimal feeding and marketing decisions

Problem definition

Modelling approach

Initial results

Further research

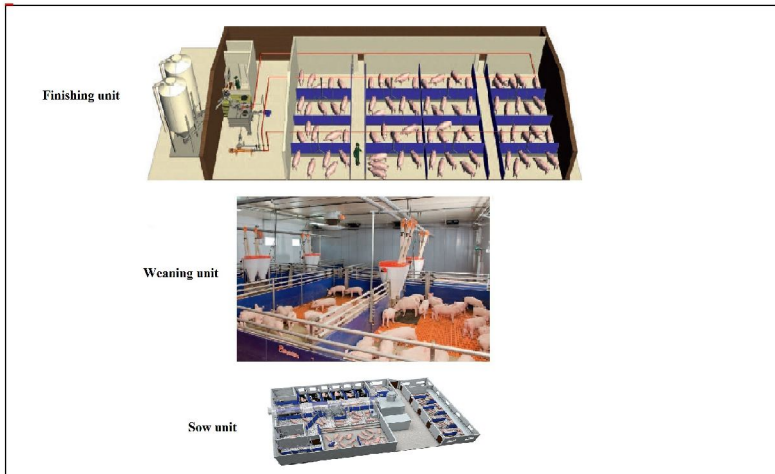
A sub-project in the work package B6

A hierarchical Markov decision process modelling feeding and marketing decisions of growing pigs

Contributions:

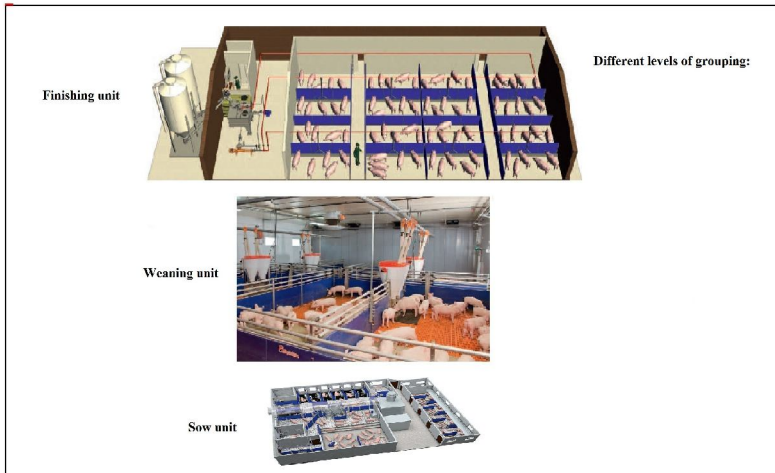
- ▶ Combination of feeding and marketing decisions by a hierarchical Markov decision process (HMDP).
- ▶ Embedding a dynamic linear model (DLM) and a dynamic generalised linear model (DGLM) into the proposed HMDP.

Problem definition

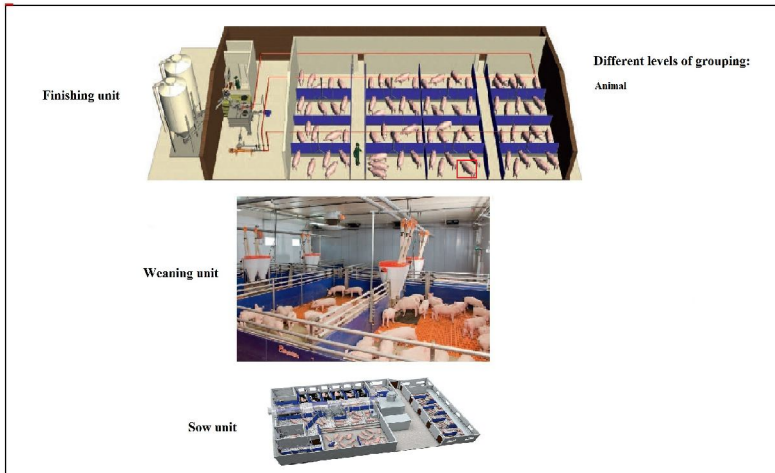


Source: <http://www.bigdutchman.de/>

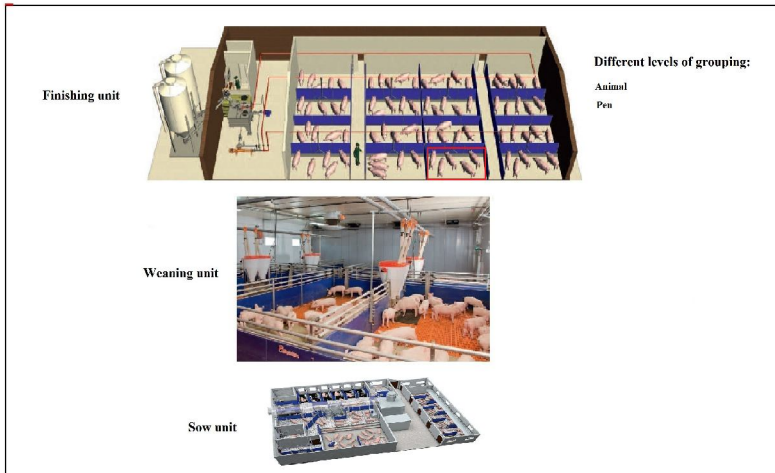
Problem definition



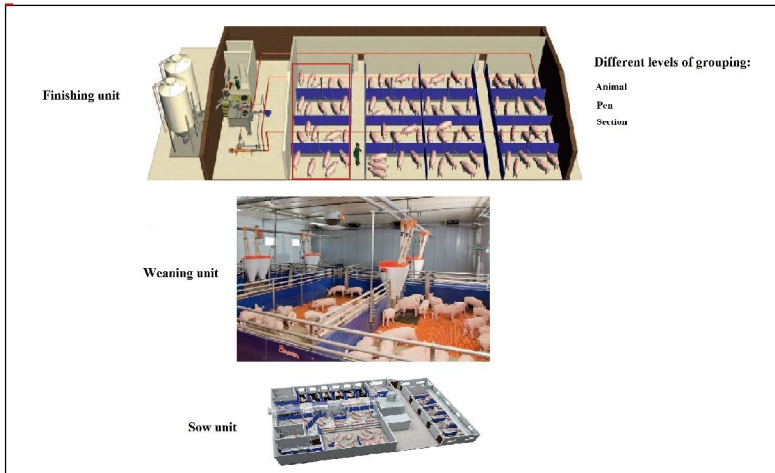
Problem definition



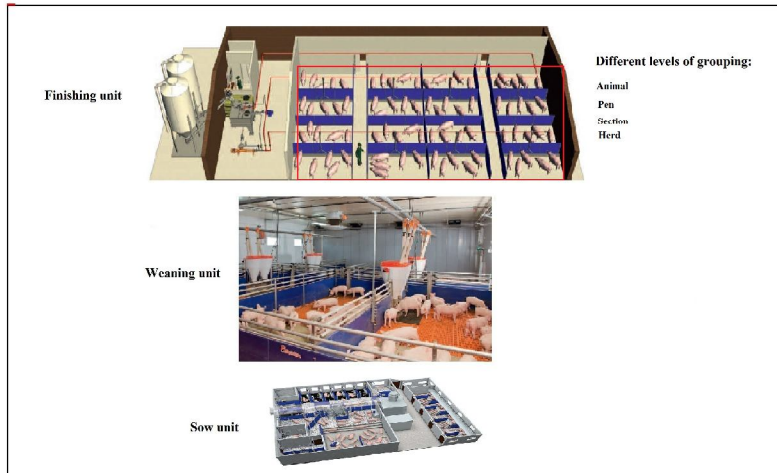
Problem definition



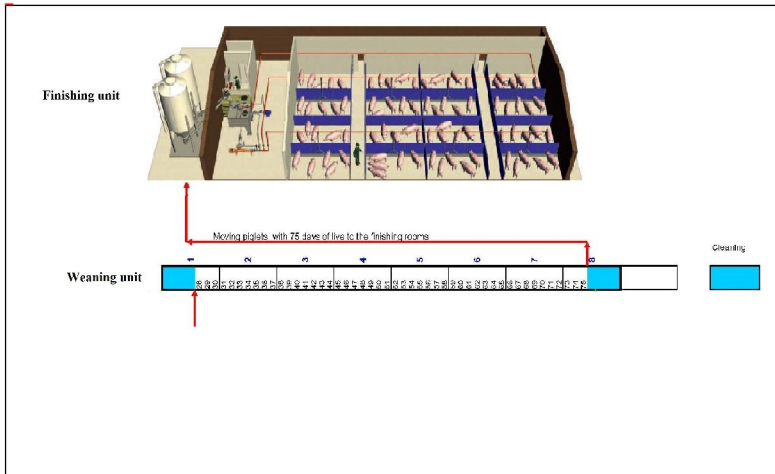
Problem definition



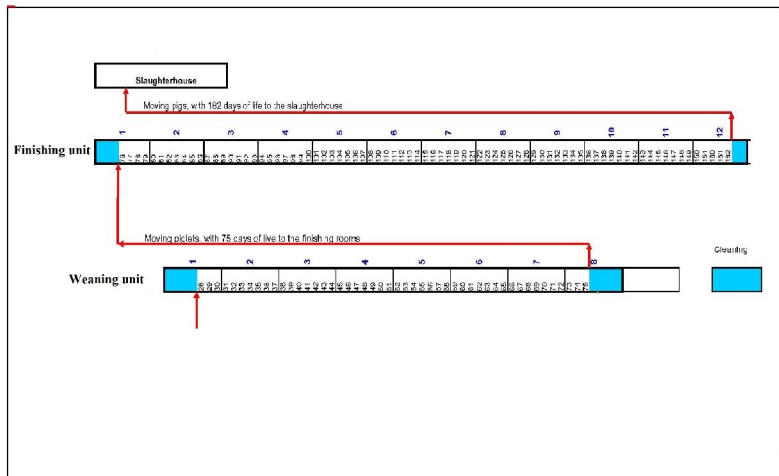
Problem definition



Problem definition

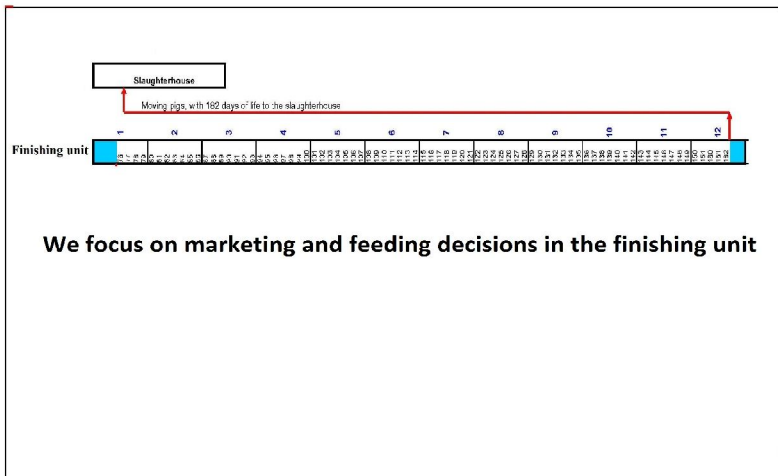


Problem definition

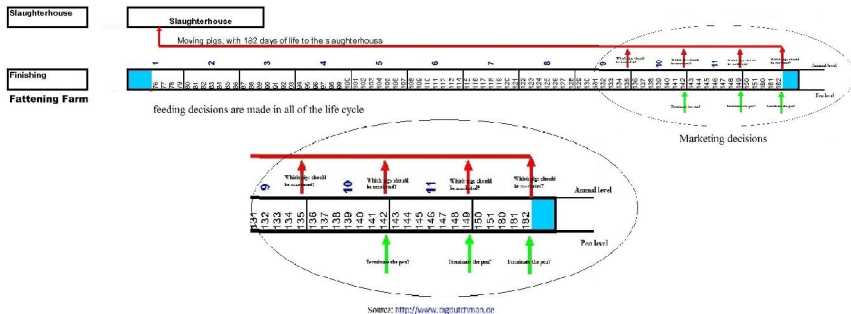


Source: <http://www.bigdutchman.de/>

Problem definition



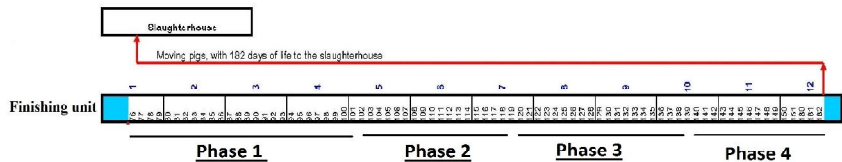
Problem definition - Marketing decisions



- ▶ Pen (or section) level: decision about when to terminate a pen and insert a new group of piglets to the pen.
- ▶ Individual (Animal) level: decisions about how to select and when to market individual animals for slaughter.

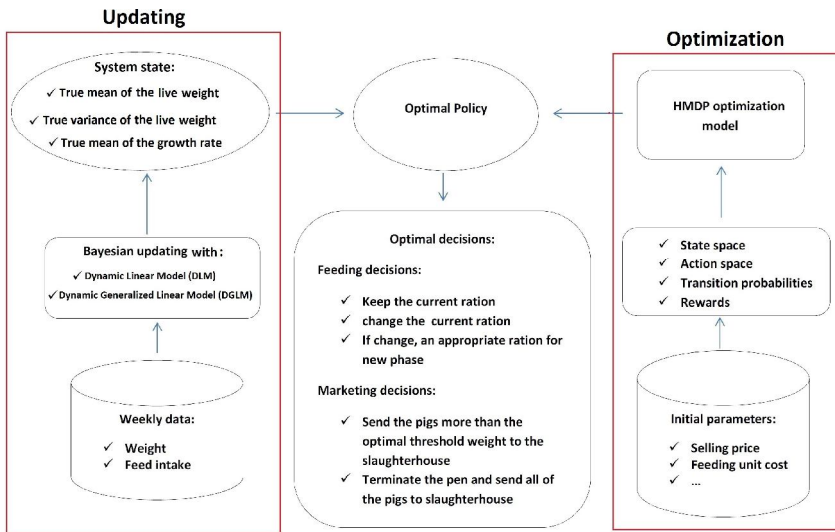
Problem definition - Feeding decisions

Phase feeding



- ▶ Decision about the time that the current ration (in the current phase) should be changed.
- ▶ Decision about finding the best new ration in the next phase.

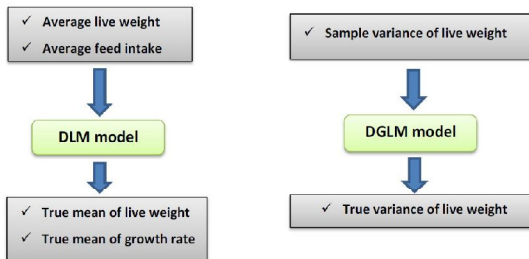
Modelling approach - framework (pen level)



Updating model - Observations

The observations in the pen for each week (or day):

- ▶ Average live weight
- ▶ Sample variance of the live weight
- ▶ Average feed intake



Updating model - DLM

- ▶ DLM is a statistical model to analyse the time series and extract the true trend of latent variables (θ_t) based on the observed data:

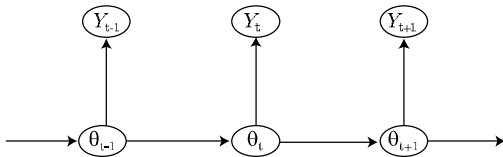


- ▶ A system equation describes the extension of the latent variable against time t:

$$\theta_t = G\theta_{t-1} + \omega_t, \quad \omega_t \sim N(0, V_t)$$

Updating model - DLM

- ▶ DLM is a statistical model to analyse the time series and extract the true trend of latent variables (θ_t) based on the observed data:



- ▶ A system equation describes the extension of the latent variable against time t:

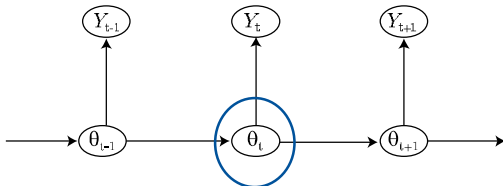
$$\theta_t = G\theta_{t-1} + \omega_t, \quad \omega_t \sim N(0, V_t)$$

- ▶ An observation equation indicates the relation between the observable (Y_t) and latent (θ_t) variables:

$$Y_t = F'\theta_t + \nu_t, \quad \nu_t \sim N(0, W_t)$$

Updating model - DLM

- ▶ DLM is a statistical model to analyse the time series and extract the true trend of latent variables (θ_t) based on the observed data:



- ▶ A system equation describes the extension of the latent variable against time t:

$$\theta_t = G\theta_{t-1} + \omega_t, \quad \omega_t \sim N(0, V_t)$$

- ▶ An observation equation indicates the relation between the observable (Y_t) and latent (θ_t) variables:

$$Y_t = F'\theta_t + \nu_t, \quad \nu_t \sim N(0, W_t)$$

Updating model - DLM

Observation equation:
$$\begin{pmatrix} W_{t,f} \\ FE_{t,f} \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ k_{1(t,f)} & k_2 \end{pmatrix} \begin{pmatrix} TW_{t,f} \\ Z_{t,f} \end{pmatrix} + v$$

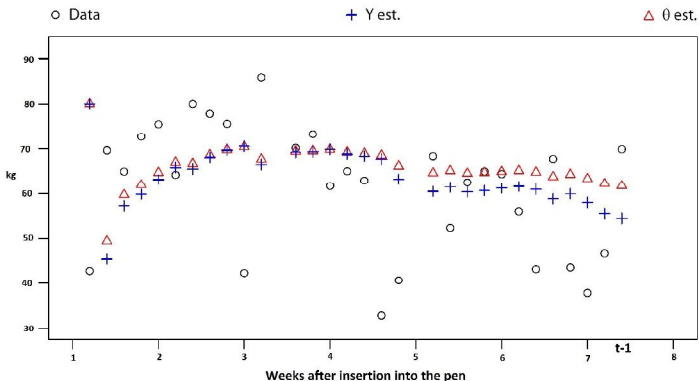
System equation:
$$\begin{pmatrix} TW_{t,f} \\ Z_{t,f} \end{pmatrix} = \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} TW_{t-1,f} \\ Z_{t-1,f} \end{pmatrix} + \omega$$

- ▶ $W_{t,f}$: Average live weight at time t
- ▶ $FE_{t,f}$: Average feed intake in interval $[t, t + 1]$
- ▶ $TW_{t,f}$: True live weight at week t
- ▶ $Z_{t,f}$: True growth rate at time t
- ▶ V : System variances (constant) - estimated by EM algorithm
- ▶ W : Observation variances (constant) - estimated by EM algorithm

Updating model - Applying Bayesian updating in the DLM

Data up to time $t - 1$: D_{t-1}

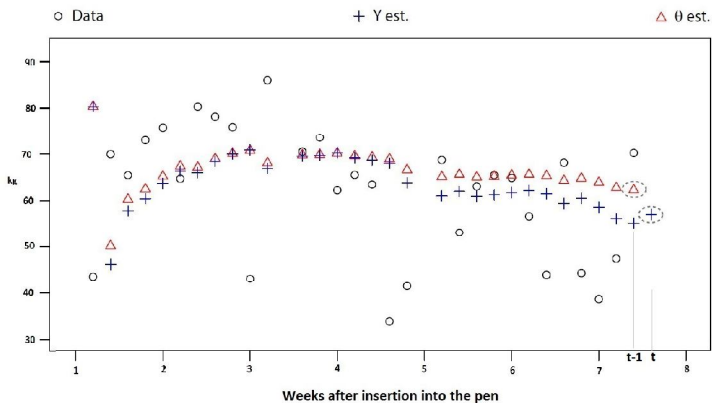
Posterior at $t - 1$: $(\theta_{t-1} | D_{t-1}) \sim N(m_{t-1}, C_{t-1})$



Updating model - Applying Bayesian updating in the DLM

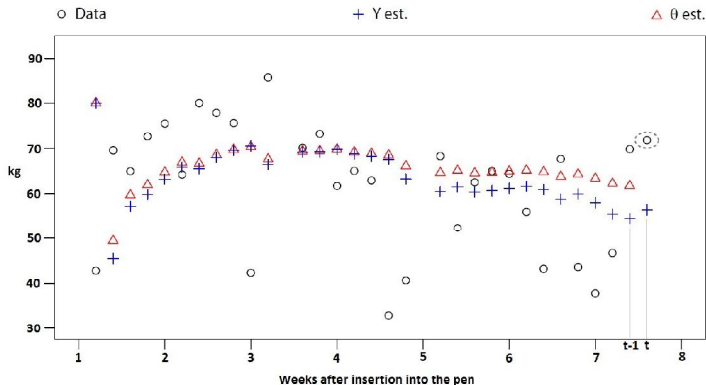
Given $(\theta_{t-1} | D_{t-1}) \sim N(m_{t-1}, C_{t-1})$ we have that

$(Y_t | D_{t-1}) \sim N(f_t, Q_t)$ where, $f_t = F_t' G_t m_{t-1}$



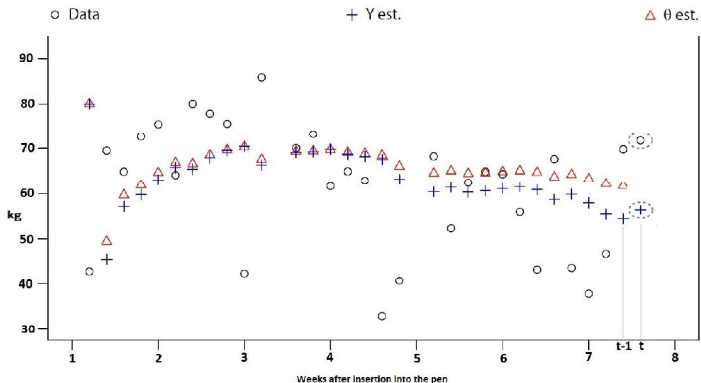
Updating model - Applying Bayesian updating in the DLM

At time t we see an observation y_t :



Updating model - Applying Bayesian updating in the DLM

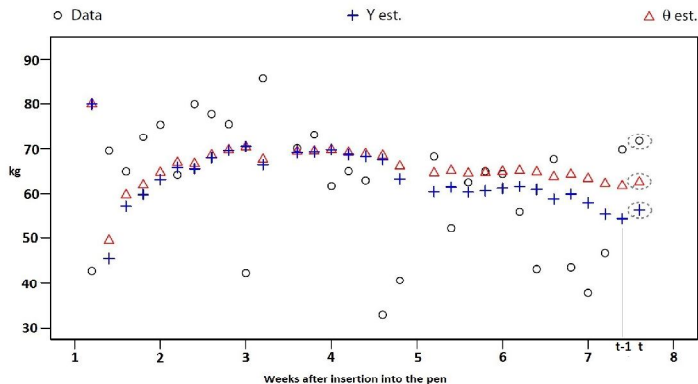
Given f_t and y_t we have that



Updating model - Applying Bayesian updating in the DLM

Given f_t and y_t we have that

$$(\theta_t \mid D_t) \sim N(m_t, C_t) \text{ where } m_t = G_t m_{t-1} + R_t F_t Q_t (y_t - f_t)$$



Updating model - DGLM

Observation equations:

$$f(s_t^2 | \eta_t) = \frac{\exp\left(-\frac{s_t^2(n_t-1)}{2\sigma_t^2}\right) \left(\frac{n_t-1}{2\sigma_t^2}\right)^{\frac{n_t-1}{2}} (s_t^2)^{\frac{n_t-3}{2}}}{\Gamma(n_t - 1)}$$

$$g(\eta_t) = \sigma_t^2$$

where $g(x) = \frac{1}{x}$

System equations:

$$\sigma_t^2 = G_t \sigma_{t-1}^2$$

- ▶ s_t^2 : Sample variance of the live weights in pen.
- ▶ σ_t^2 : True variance of the live weight in the pen.
- ▶ $G_t = \frac{t}{(t-1)}$
- ▶ n_t : Number of pigs in the sample for s_t^2

Updating model - Validating the DLM

- ▶ To validate the DLM model, the observations were generated by a simulation approach and we supposed the true means are given in this step.
- ▶ The filtered data (output of the DLM model) show a good prediction of the true means of the live weight and the true means of the growth rate in the pen.
- ▶ Initial results indicate a good learning capability of the suggested DLM model.

