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## Modelling impacts of performance on the probability of reproducing, and thereby on productive lifespan, allow prediction of lifetime efficiency in dairy cows

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Reproductive success is a key component of lifetime efficiency — which is the ratio of energy in milk (MJ) to energy intake (MJ) over the lifespan, of cows. At the animal level, breeding and feeding management can substantially impact milk yield, body condition and energy balance of cows, which are known as major contributors to reproductive failure in dairy cattle. This study extended an existing lifetime performance model to incorporate the impacts that performance changes due to changing breeding and feeding strategies have on the probability of reproducing and thereby on the productive lifespan, and thus allow the prediction of a cow's lifetime efficiency. The model is dynamic and stochastic, with an individual cow being the unit modelled and one day being the unit of time. To evaluate the model, data from a French study including Holstein and Normande cows fed high-concentrate diets and data from a Scottish study including Holstein cows selected for high and average genetic merit for fat plus protein that were fed high- v. low-concentrate diets were used. Generally, the model consistently simulated productive and reproductive performance of various genotypes of cows across feeding systems. In the French data, the model adequately simulated the reproductive performance of Holsteins but significantly under-predicted that of Normande cows. In the Scottish data, conception to first service was comparably simulated, whereas interval traits were slightly under-predicted. Selection for greater milk production impaired the reproductive performance and lifespan but not lifetime efficiency. The definition of lifetime efficiency used in this model did not include associated costs or herd-level effects. Further works should include such economic indicators to allow more accurate simulation of lifetime profitability in different production scenarios.

Keywords: dairy cow, lifetime efficiency, productive lifespan, reproduction, nutrient partitioning

### **Implications**

This paper extended an existing lifetime performance model to incorporate the impacts that varying milk yield, energy balance and body condition have on the probability of reproducing and the productive lifespan, thereby allowing prediction of a cow's lifetime efficiency. The model adequately simulated the productive and reproductive performance of various genotypes of cows across feeding systems. Thus, it can be used to study the effect of future selection and management on animal performance and efficiency.

### Introduction

It is widely accepted that single-trait selection for greater milk production in dairy cattle is associated with significant declines in animal health (Berry et al., 2011) and reproductive performance (Albarrán-Portillo and Pollott, 2013). These problems not only impair the lifetime production efficiency of dairy cows via reduced longevity by increasing involuntary culling (Seegers et al., 1998), but also increase veterinary and reproductive costs (Hogeveen et al., 2011) and environmental losses (Garnsworthy, 2004), and affect animal welfare (Oltenacu and Algers, 2005). An increased milk yield, a negative energy balance and a low body condition have been considered the major biological contributors to the failure of reproduction and thus to a curtailed productive lifespan of dairy cows (Friggens et al., 2010). However, the individual effects of milk yield, energy balance and body condition on reproduction have been difficult to quantify because they are often correlated. Experiments specifically designed to break these correlations are unfortunately rare (Wright et al., 1992), and a modelling approach to integrate

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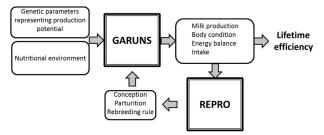
available data seems necessary (Blanc *et al.*, 2001). Management strategies such as breeding (e.g. selecting for high *v.* low genetic merit cows) and feeding programmes (e.g. high- *v.* low-energy diets) can alter milk yield and body reserves, and consequently reproductive performance (Cutullic *et al.*, 2011). In this context, if the objective is to predict lifetime efficiency (LTE) there is a need to be able to predict responses of dairy cows in terms of both productive and reproductive performance under different management strategies. Such a predictive capability is especially important in the case of dairy production systems where wide variation might exist between animal genotypes and between the environments in which they are managed.

Individual-based herd models that describe each animal can be used to simulate animal responses to management practices (Brun-Lafleur et al., 2013). Unfortunately, in most of these models, the reproductive process is represented by assuming a fixed reproductive efficiency such as fixed calving to ovulation or fixed parturition to conception intervals (Blanc and Agabriel, 2008; Martin and Sauvant, 2010). Although recent work by Inchaisri et al. (2010) and Brun-Lafleur et al. (2013) has recognized the possible influences of milk yield and body reserves on reproductive performance, the simulation of these traits is only based on genetic potential and physiological stages, without the possibility of including an interaction with feeding regimes. Additionally, these models were only developed for a single lactation and thus cannot be used to predict the lifetime performance and LTE of animals.

The objective of this study is to extend the published lifetime nutrient partitioning model of Martin and Sauvant (2010), which predicts the LTE of individual dairy cows under various management strategies, by incorporating the potential influences of animal performance on reproductive performance. It is hypothesized that a high milk yield, a negative energy balance and a low body condition negatively affect reproductive performance, and thereby the productive lifespan, which is expected to have consequences for the LTE of dairy cows.

#### Material and methods

The model used in this study is dynamic and stochastic, with the individual cow being the unit modelled and one day being the time unit used within the model. The lifetime of a cow is considered from birth until culling. A schematic presentation of this model is presented in Figure 1. The core of this model relies on the lifetime nutrient partitioning model of Martin and Sauvant (2010), referred to here as GARUNS, which simulates lifetime performance of individual cows. The inputs of GARUNS include genetic scaling parameters (specified in next section), feeding information (i.e. dietary energy density), reproduction (i.e. conception and parturition) and management (i.e. culling) events. The outputs of GARUNS are daily feed intake, milk yield and composition, foetal growth, calf birth weight, BW and composition, body condition score (BCS) and energy balance

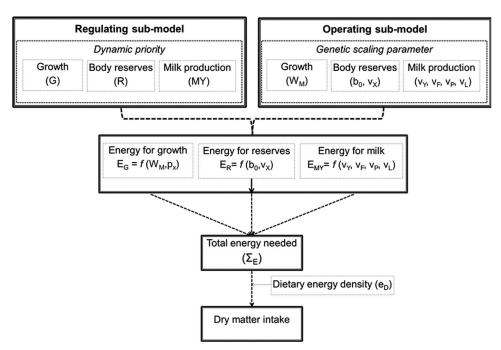


**Figure 1** Schematic presentation of the model for predicting lifetime efficiency of individual cows that includes the GARUNS model.

throughout the lifespan of a cow. The reproduction sub-model developed in this study uses the predictions of milk yield, BCS and energy balance from GARUNS as inputs to simulate reproductive responses in terms of conception rates, and thus timing of conception and parturition, and culling decisions, which are in turn transferred back to GARUNS. In the present study, the lifespan of individual cows is mainly affected by reproductive performance and culling policy. The effect of diseases — for example, mastitis or fatty liver — on the survivability of cows is not currently incorporated.

Description of lifetime nutrient partitioning model (GARUNS) The GARUNS model consists of a regulating sub-model providing the driving force to control the function of an operating sub-model. An adapted schematic of GARUNS is presented in Figure 2. The regulating sub-model describes the dynamic partitioning of a female mammal's priority between life functions – growth (G), ageing (A), regaining of body reserves (R), and energy supply of the unborn (U), newborn (N) and suckling (S) calf - over her lifespan. This dynamic pattern of relative priorities is assumed to be general for all individual cows and always sums to one. The operating sub-model uses the relative priorities to partition energy intake between foetal growth, BW and composition, milk yield and composition during repeated reproductive cycles and subsequently over the cow's lifespan. Genetic scaling parameters are incorporated into the operating sub-model to scale the individual performance potentials of, for example, weight or milk production. It is noted that these genetic scaling parameters are not directly related to the widely reported breeding values commonly published in the animal genetics literature. They act as multipliers on the different dynamic priorities and thus provide the means to create different levels of genetic potential for different cows.

In this context, the variation in genetic scaling parameters describes the variation in genetic potential between animals for the different GARUNS life functions. In the model, an animal is described in terms of its genotype for growth, capacity to store and mobilize reserves, and milk production using the following genetic scaling parameters: non-labile body mass at maturity ( $W_M$ ), daily rate of regaining body reserves ( $b_0$ ), labile body mass mobilization index ( $\nu_\chi$ ), peak milk yield potential ( $\nu_\gamma$ ), milk fat secretion ( $\nu_F$ ), milk protein secretion ( $\nu_P$ ) and milk lactose secretion ( $\nu_L$ ).  $W_M$  is given in kg, whereas the remaining parameters are fractions.



**Figure 2** Schematic diagram of the GARUNS model (reprinted from Phuong *et al.* (2015)), showing how energy requirements for different life functions (growth, body reserves and milk) are derived from genetic scaling parameters and lifetime profiles of relative priorities for the different life functions. The priorities for foetal growth, ageing and energy needed for foetal growth are not presented as they are assumed to be common trajectories for every individual.

Thus, to describe each genotype a specific set of these genetic scaling parameters is needed. External triggers of successful insemination (i.e. conception) drive the changes between non-reproductive and reproductive states of the animal, cueing in the dynamic priority trajectories described in the regulating sub-model.

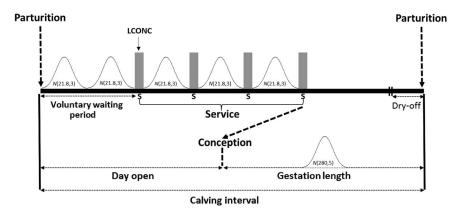
As shown in Figure 2, the interactions between dynamic priority and genetic scaling parameters make it possible to quantify the total amount of energy required to fulfil all functions ( $\Sigma_E$ ). The resulting dry matter intake (DMI) is quantified by multiplying  $\Sigma_E$  with the predetermined dietary energy density ( $e_D$ ). The coefficients for energy partitioning to different life functions are  $E_G/\Sigma_E$ ,  $E_R/\Sigma_E$  and  $E_{MY}/\Sigma_E$ , where  $E_G$ ,  $E_R$  and  $E_{MY}$  are energy required for growth, reserves storage and milk production, respectively.

### Description of the reproduction sub-model

Owing to the main interest of this study being the LTE of individual cows, for which a key component is lifespan of the animals, the only reproductive information required is that which creates plausible variation in lifespan. This can simply be seen as a result of whether or not, and when, the cow becomes pregnant, and calves. Instead of modelling all the individual steps of the reproductive process (uterine evolution, resumption of oestrus cycle, oestrus expression, conception and embryonic survival), the likelihood of conception of a viable calf (LCONC) as the final product of the reproductive process is assumed. This is the accumulative result of other steps involved towards having a newborn calf. LCONC is assumed to be affected by milk yield, energy

balance and body condition of the cow (Friggens et al., 2010). The influences of parity (primiparous  $\nu$ . multiparous), oestrus number and disease on LCONC are not explicitly incorporated. The key events in the reproduction sub-model include conception and parturition. Figure 3 shows that, after parturition, cows are assumed to have a period of postpartum anoestrus of  $21.8 \pm 3$  days (Darwash et al., 1997). Thereafter the cow is assumed to have an oestrus of 3 days, which takes place  $21.8 \pm 3$  days after the previous oestrus period. The breeding period commences after a voluntary waiting period, set to 50 days (Löf et al., 2012). Before the end of the voluntary waiting period and outwith the 3-day oestrus window, LCONC is assumed to be zero. During the breeding period, artificial insemination is assumed to take place at every oestrus with a LCONC greater than a threshold value (0.4) to reflect the idea that weak oestruses will be poorly expressed and thus not detected by the inseminators. Insemination is simulated using a random draw from a uniform distribution, U(0, 1). A cow is assumed pregnant if the insemination value is smaller than the LCONC value.

LCONC is calculated using components determined by milk yield (LCONC<sub>MY</sub>), energy balance (LCONC<sub>EB</sub>) and body condition (LCONC<sub>BCS</sub>) of the cows (equation (1)). Under perfect conditions, these LCONC components each have a value of 1 that can be driven down to 0 as described below. A maximum LCONC value of  $0.6 \pm 0.06$  is assumed, which is supported by reported variation of 57% to 63% for the conception rate of Holstein virgin heifers (Kuhn *et al.*, 2006). This assumption is made in light of our initial hypothesis that fertility is mainly affected by high milk production, a negative energy balance and a low body condition, which do not exist



**Figure 3** Representation of the reproduction sub-model, showing the assumed intervals for *postpartum* anoestrus, oestrus cycles, and pregnancy, and their variability (as standard deviations). The grey bars indicate when the likelihood of conception is not zero – that is, in the 3 days around oestrus. The time axis is not to scale as this is a visualization of the model.

in virgin heifers. Consequently, LCONC is computed as follows:

$$LCONC = 0.60 \times (LCONC_{MY} \times LCONC_{EB} \times LCONC_{BCS})$$
 (1)

Gestation length is assumed to follow a normal distribution, *N* (280, 5) days (Hansen *et al.*, 2004). Factors influencing gestation length such as breed, age or birth weight of the calf were not considered.

Effect of milk production on the likelihood of conception The underlying mechanism for the negative effect of milk yield (MY) on the fertility of dairy cows was investigated by Wiltbank et al. (2006). As high milk yield is supported by high dry matter intake and consequently high blood flow in the liver, it increases the speed of clearance of oestradiol and progesterone. The low concentration of plasma oestradiol reduces the stimulation of the centres that control oestrus behaviour, and thus the behavioural expression of oestrus, whereas a reduced circulating concentration of progesterone is associated with a low pregnancy rate (McNeill et al., 2006). In this study, it is assumed that as soon as the cows are producing milk it has a negative effect on LCONC; as such, LCONC decreases in a sigmoidal manner with increasing MY. The likelihood of conception resulting from the effect of milk yield (LCONC<sub>MY</sub>) is quantified in equation (2).

$$LCONC_{MY} = 2/(1 + exp(Rat_{MY} \times MY))$$
 (2)

where MY is daily milk yield (kg/day) and  $Rat_{MY}$  is a constant for the extent of the effect of milk yield on the likelihood of conception. The value of  $Rat_{MY}$  (0.012) was estimated from the study by Cutullic *et al.* (2011) where the experiment was designed to separate the effect of MY on reproductive performance from the effects of energy balance and body condition.

Effect of energy balance on the likelihood of conception The effect of negative energy balance (NEB) is commonly attributed to the suppression of the pulsatility of luteinizing hormone (LH) and reductions in the ovarian responsiveness to LH stimulation (Butler, 2000). During the period of

negative energy balance, plasma glucose, insulin and insulinlike growth factor-I (IGF-I) are reduced. Insulin is known to stimulate bovine follicular cells (Spicer *et al.*, 1993), whereas IGF-I is critical to ovarian follicular development (Butler *et al.*, 2006). Consequently, NEB shifts the course of *postpartum* ovarian activity and strongly influences the resumption of ovarian cycles (Senatore *et al.*, 2010). Other authors have also reported an inferior oocyte quality in negativeenergy-balance cows (Leroy *et al.*, 2008). The effect of NEB on the likelihood of conception (LCONC<sub>EB</sub>) is formalized in equation 3.

$$LCONC_{EB} = 2/(1 + exp(-Rat_{EB} \times EB)) \tag{3}$$

where EB is daily energy balance (MJ/day) and  $Rat_{EB}$  is a constant parameter. The value of  $Rat_{EB}$  (0.027) was derived from the study by Wright *et al.* (1992). This experiment on beef cattle was designed to break the correlation between body fatness and body mobilization by applying two feeding levels during lactation to cows that had previously been made thin or fat. Thus, it is possible to separate the effect of body condition and NEB on reproductive performance, if no effect of MY is assumed. Energy balance is computed from GARUNS outputs as follows:

$$EB = E_{Intake} - (E_{Growth} + E_{Maintenance} + E_{MY} + E_{Pregnancy})$$
 (4)

where EB is energy balance (MJ/day);  $E_{Intake}$  is energy intake (MJ/day); and  $E_{Growth}$ ,  $E_{Maintenance}$ ,  $E_{MY}$  and  $E_{Pregnancy}$  are energy requirements for growth, maintenance, milk production and pregnancy, respectively, expressed as metabolizable energy (MJ/day).

Effect of body condition on the likelihood of conception In the literature, depressed reproductive performance in thin compared with fat cows has been observed (Wright *et al.*, 1992; Pryce *et al.*, 2001). Low body condition is related to low level of leptin, which is also linked to reproductive function (Spicer, 2001). The effect of body condition on the likelihood of conception (LCONC<sub>BCS</sub>) is formalized in equation 5. To measure the body fatness of cows, BCS is used on a scale of 0 to 5 (0 = emaciated and 5 = extremely fat).

It is assumed that a level of BCS  $\geqslant$  3 has no effect on LCONC<sub>BCS</sub>.

$$LCONC_{BCS} = 2/(1 + exp(-Rat_{BCS} \times (BCS-3)))$$
 (5)

where  $Rat_{BCS}$  is a constant parameter. The value  $Rat_{BCS}$  (0.425) was obtained from the paper by Wright *et al.* (1992).

#### Culling policy

The reproductive culling rules assumed for this study are as follows. Cows are culled if they fail to become pregnant after four inseminations and MY on that day is <25 kg/day. If the insemination number is 4 but cows are still producing ≥25 kg/day of milk, a fifth insemination is allowed; if this does not result in pregnancy they are culled. To take into account the fact that some cows might not express oestrus for a long time, another culling rule is included where cows are culled anyway if they are not pregnant after 160 days postpartum. For simplicity, at this stage, the effects of management factors such as age or population structure on culling decisions were not considered.

#### Model implementation and evaluation

The model was implemented in C++ programing language on the RECORD platform (Bergez et al., 2013). To evaluate the model, two data sets were taken from the literature. In the first data set of Cutullic et al. (2011) and Delaby (personal communication, 2014), recorded at the Le Pin research farm, two breeds of cows (Holstein Friesian selected for MY and dual-purpose Normande cows) fed a high-concentrate diet were used to evaluate the effect of genetic selection on reproductive performance and LTE. The second data set was taken from the long-term experiment carried out at the Langhill Dairy Cattle Research Centre (Pryce et al., 1999; Coffey et al., 2004). In the Langhill data, high- and low-feeding systems were applied to two lines of Holstein Friesian cows: a control line and the line that had been selected for kilograms of fat plus protein yield. This allowed testing of the module under a genotype and feeding system interaction. The voluntary waiting period was specifically set to 65 days for the Langhill cows to match the data (Smith; personal communication, 2014).

GARUNS was first adjusted to simulate populations of cows that match the observed productive performance in the selected data sets. This was done using the genetic scaling parameters representing milk production level ( $\nu_y$ ), mature weight, ( $W_M$ ), daily rate of regaining body reserves ( $b_0$ ) and labile body mass mobilization index ( $\nu_x$ ). The higher the values of  $\nu_y$  and  $W_M$ , for example, the higher the capability of producing milk and the larger the cows. All cows were assumed to share the same capability of secreting milk fat ( $\nu_F$ ), milk protein ( $\nu_P$ ) and milk lactose ( $\nu_L$ ), and thus milk composition. The within-population variability in  $\nu_y$ ,  $W_M$ ,  $b_0$  and  $\nu_X$  was simulated using the standard deviations and correlations reported by Phuong *et al.* (2015).

Low feeding was simulated using a level of dietary energy density (e<sub>D</sub>) to reduce the total energy intake of cows, and thus performance (i.e. milk yield and BW), to the level

reported in the data, whereas high feeding implies no limitation to energy intake. Having created the same levels of milk production, energy balance and body condition as observed in the data sets, the LCONC module was then used to estimate the resulting reproductive performance and productive lifespan. Simulated results for conception to first service (FSC), days from parturition to first service (DFS), days from parturition to conception or days open (DOP), and calving interval (CI) were then compared with those observed in the experimental data set. To test the repeatability of the model, each scenario was run 20 times, with 200 cows each. To evaluate the reliability of the reproduction module in predicting the reproductive performance of cows, the likelihood of the mean simulated value being part of the distribution of observed values reported in the experiment was calculated using the equation suggested by Friggens et al. (2007):

likelihood = 
$$1 - F[(x-\mu)/\sigma]$$
 (6)

where x is the mean simulated value,  $\mu$  is the mean of the observed value,  $\sigma$  is the standard deviation of the observed value, and F is the distribution function for the standard normal distribution.

Welch's *t*-test was also performed using R software version 3.2.1 (R Development Core Team, 2015) for comparing the means of observed and simulated values.

#### Calculation of LTE

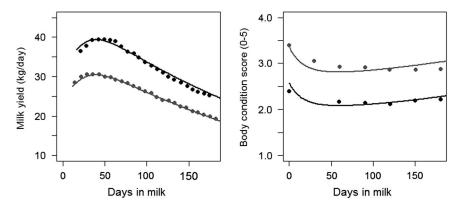
The LTE of individual cows in each scenario was calculated using the following equation:

$$LTE = \frac{\sum_{1}^{N} E_{Milk}}{\sum_{1}^{N} E_{Intake}} \times 100.$$
 (7)

where LTE is the lifetime efficiency (%),  $E_{Milk}$  is the total energy in milk (MJ),  $E_{Intake}$  is the total energy intake (MJ) and N is the time from birth to culling (day).

#### **Results**

An example simulation of two genotypes of cows differing in milk production level and BCS is given in Figure 4, using the genetic scaling parameters  $\nu_y$ ,  $W_M$ ,  $b_0$  and  $\nu_x$ . In this example, the high MY and low BCS curves represent the average Holstein cow fed a high-concentrate diet in Cutullic *et al.* (2011), whereas the low MY and high BCS curves represent the average Normande cow. Figure 4 shows that the GARUNS model adequately simulated lactation and BCS curves of the average Holstein and Normande cows. No comparison in BW curves could be made because of insufficient data, but the simulated live weights 1 week *postpartum* of 678 and 700 kg for Holstein and Normande cows, respectively, were comparable to the reported data. More generally, different populations of cows differing in genetic potential — for instance, high-  $\nu$ . low-milk-producing cows — could be simulated by scaling the



**Figure 4** Model simulation of milk yield and body condition score of Holstein and Normande cows in the Le Pin data. Circles are observed data set and lines are simulated outputs. Black represents Holstein cows and grey represents Normande cows.

**Table 1** Simulated effect of high v. low genetic merit for milk production on reproductive performance and LTE compared with data from Le Pin cows (mean  $\pm$  standard deviation<sup>1</sup>)

		Holstein			Normande	
Trait	Observed	Simulated	<i>P</i> -value	Observed	Simulated	<i>P</i> -value
FSC	$38 \pm 5 (45)^2$	41 ± 5	0.37	58 ± 7 (36)	48 ± 4	<0.05
DFS	$80 \pm 14 (40)$	$84 \pm 30$	0.12	$75 \pm 14 (26)$	$67 \pm 15$	< 0.05
DOP	$97 \pm 39 (49)$	$98 \pm 31$	0.81	$92 \pm 39 (40)$	$85 \pm 26$	0.07
CI	$377 \pm 34 (49)$	$379 \pm 31$	0.60	$376 \pm 34 (36)$	$367 \pm 26$	< 0.05
LSP	NA	$1762 \pm 911$	NA	NA	$2875 \pm 1606$	NA
LTE	NA	$45 \pm 7$	NA	NA	$41 \pm 4$	NA

FSC = first service conception (%); DFS = days from calving to first service (d); DOP = days from calving to conception or days open (d); CI = calving interval (d); LSP = lifespan (d); LTE = lifetime efficiency (%); NA = not available.

Sample size of n = 5 for FSC in observed and simulated populations, n = 200, 110 and 133 for remaining traits in simulated, observed Holstein, and observed Normande populations, respectively.

Values of standard deviation for the observed reproductive traits from the Le Pin cows were assumed from the study of Pryce et al. (1999).

genetic scaling parameters up and down. The performance of control and selected cows from the Langhill data were also simulated in the same way (data not shown).

Effect of genotypes on reproductive performance and LTE The simulated reproductive performance, lifespan and LTE of the two genotypes of cows in the example of Figure 4 are presented in Table 1. The model satisfactorily predicted reproductive performance of the Holstein cows (P > 0.05) but considerably under-predicted that of the Normande cows (P < 0.05). Table 1 gives the likelihood of a given mean simulated value of reproductive traits being part of the observed population. For example, in the Le Pin Holstein cow data the likelihood of a mean simulated FSC of 41 being part of the observed population was 45%, indicating that in the population of observed FSC values reported by Cutullic *et al.* (2011) 45% of them would be >41%.

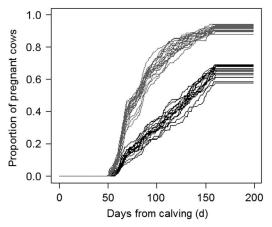
Cows with high genetic merit for milk production (Holstein) were predicted to have a more inferior reproductive performance than were low genetic merit ones (Normande) — lower conception rate, higher days to first service, higher days open, longer calving interval and shorter

lifespan. Despite having a worse reproductive efficiency and shorter productive longevity, high-milk-producing cows still have higher LTE than the low-milk-producing ones. The proportion of cows that become pregnant by a given number of days in milk is lower for the simulated Holstein cows compared with Normande cows (Figure 5). Table 1 also shows a high consistency between the simulations with low standard errors that can simply be computed by dividing standard deviation by squared-root of sample size.

# Effect of genotypes and feeding levels on reproductive efficiency and LTE

Table 2 presents simulated reproductive traits in comparison with the corresponding records in the Langhill data. The simulated FSC were comparable to those of the observed data with the likelihood of being part of the observed population ranging from 41% to 49% (Table 2). DFS are on average 6 days longer in the simulated results compared with the data records, whereas the simulated DOP and CI are on average 10 days shorter than the observed values. Generally, cows selected for maximal milk production had lower reproductive performance, and thus a shorter lifespan.

<sup>&</sup>lt;sup>2</sup>Values reported in the parentheses indicate the likelihood of the mean simulated value of that trait being part of the corresponding population of observed values (%).



**Figure 5** Simulated proportion of pregnant cows by days in milk of Holstein (black lines) *v.* Normande (grey line) cows from the Le Pin data set. Each line is one repetition of a simulation with a population of 200 cows.

However, the LTE of individual cows was mainly driven by milk production level rather than reproductive performance and lifespan, which might explain why the simulated LTE of the selected cows was higher than that of the control cows. No clear effect of level of concentrate feeding on reproductive performance could be observed in the simulation. These results are confirmed by the plots of the simulated proportion of cows becoming pregnant for control and selected populations in high-  $\nu$ . low-concentrate systems presented in Figure 6.

#### **Discussion**

The main purpose of this study was to incorporate the impact that performance level has on the reproductive success of individual cows into an existing lifetime performance model, and thereby allow prediction of cow LTE. The likelihood of conception of a viable calf as the final product of the reproductive process was made sensitive to performance levels – namely, milk yield, body condition and energy balance. This allowed the model to evaluate at each time-step within the relevant phases of lactation whether or not the cow becomes pregnant and, if so, when. On the basis of these reproductive outcomes the model was able to include variation in productive lifespan within the simulations of lifetime performance and LTE. This model development was evaluated using two data sets with varying performance levels. To simulate different genotypes of cows (high genetic merit, i.e. selected, compared with low genetic merit, i.e. control, cows), genetic scaling parameters were used to create differences in production levels such as milk yield, growth capacity or mature weight. Thus, the testing reported in the present study is only of the reproductive module (GARUNS has been tested in Martin and Sauvant (2010) and Phuong et al. (2015)).

Comparison of simulation outputs of reproductive performance with observed data

Generally, the model showed a good ability to consistently simulate the reproductive performance of various genotypes

of cows across nutritional environments as the predicted values are close to the observed ones, and changes are in the same direction between breeds. For the Le Pin data, the model under-predicted the reproductive performance of Normande cows but the simulated outputs were comparable with the records of Cutullic *et al.* (2011) for Holstein cows. The under-prediction of the reproductive performance of Normande cows was probably due to the fact that the constants in the present reproduction sub-model – that is Rat<sub>MY</sub>, Rat<sub>EB</sub> and Rat<sub>BCS</sub> – were derived from the studies including Holstein cows. Indeed, there have been reports of significant innate variation between breeds for reproductive performance even when fed the same diet (Berry *et al.*, 2003; Cutullic *et al.*, 2011).

Compared with the Langhill data the model adequately predicted FSC while slightly over- and under-predicted DFS, and DOP and CI, respectively. However, such discrepancies between simulated and observed values were rather small – that is, 5 days for DFS and 10 days for DOP and CI, on average – and thus might not be practically important.

It is particularly encouraging that the model was able to predict the direction and magnitude of differences in reproductive performance of cows when comparing control  $\nu$ . selection line cows, and high- $\nu$ . low-concentrate feeding. High-yielding cows took longer to be inseminated or to become pregnant and thus had a longer CI than the low-milk-producing individuals. Although the FSC was not different between the two genotypes, the overall conception rate, or the proportion of cows becoming pregnant (Figure 6), and the lifespan (Table 2) were reduced in the high-genetic-merit (i.e. selected) compared with the low-genetic-merit (i.e. control) cows. These results are consistent with literature comparisons of differences in genetic merit for milk production (Horan *et al.*, 2005; Yániz *et al.*, 2008).

When examining the effect of the feeding system, the simulated outputs show that there were no clear differences in reproductive performance, lifespan and LTE between the low- and the high-concentrate-feeding groups. The plots (not shown) of simulated effects of milk production, BCS and energy balance on the likelihood of conception - that is, LCONC<sub>MY</sub>, LCONC<sub>BCS</sub> and LCONC<sub>EB</sub>, respectively – of individual cows in high- v. low-concentrate-feeding groups indicate that the likelihood of conception in the low-feeding group was reduced via low BCS and negative energy balance but this was compensated for by lower milk production, whereas the major negative effect of high feeding was on the likelihood of conception due to high milk production. These results agree with the conclusion of Cutullic et al. (2011) that low-fed cows, on the one hand, may suffer from greater body condition loss and poorer body condition but on the other hand may benefit from a lower level of milk production. Interestingly, when examining the observed feed effects, Pryce et al. (1999) found little overall difference in the reproductive performance between animals fed a high v. a low concentrate. However, working on the same data at Langhill but with a division of records into heifers and

multiparous cows over two lactations, McGowan et al. (1996) reported that low-concentrate-fed heifers had a better reproductive performance but found the opposite pattern with multiparous cows. In the data of Pryce et al. (1999) 34% of records were from heifers. Further to this, the effect of high- v. low-concentrate feeding (or dietary energy density) is not consistent in the literature: there are studies that have reported better reproduction when fed low-concentrate compared with high-concentrate diets (Horan et al., 2005; Walsh et al., 2008). There are also studies showing the opposite effect (Pedernera et al., 2008; Wall, 2008). Cutullic et al. (2011) showed a lower proportion of cows being inseminated during the breeding period when fed a low-energy-density diet compared with a control diet. This indicates the necessity of carefully considering the difference between heifers and cows in their capability to reproduce, especially the way they respond to a change in the nutritional environment.

## Effect of genotype and feeding system on lifespan and LTE of individual cows

In the present study, the simulated total lifespans of Holstein cows in either control or selected lines fed high-concentrate diets are comparable to reports from the literature; that is, the simulated lifespan varies from 2225 to 2856 days, whereas it is ~2200 days for British Holsteins (Pritchard et al., 2013) and 2128 days for Dutch Holsteins (CRV, 2014). In this model it was assumed that all cows start breeding at 450 days of age, and because the gestation length is on average 280 days the start of first lactation was on average at 730 days. Given this, the productive lifespan (time from first calving to culling) was calculated to be on average 1810 days or 4.4 lactations (productive lifespan/calving interval). This is higher than the reported mean productive lives of 3.3 lactations for Austrian Holstein cows (Knaus, 2009), 1500 days for British Holstein cows (Pritchard et al., 2013) and three lactations for American Holstein (Hare et al., 2006). The longer simulated lifespan might be attributed to the fact that in this study the lifespan of individual cows is mainly driven by reproductive performance and a simple culling policy. The effect of diseases – for example, mastitis or fatty liver - on cow survivability is not currently incorporated.

The simulation outputs showed that selection for greater milk production decreases the reproductive performance and lifespan (Table 2). Interestingly, in terms of efficiency, the present model suggests that such decreases in reproductive performance and lifespan are probably compensated for by high milk production, which explains the slightly higher LTE of selected cows compared with control cows (48% v. 45%, Table 2). In this study, a simple definition of LTE was used – the within-cow ratio of total energy captured in milk over total energy intake – and thus the improvement of LTE relies on the principle of diluting maintenance requirements by increasing milk production (VandeHaar and St-Pierre, 2006) or reducing the number of unproductive days (Lormore and Galligan, 2001). At the animal level, milk production had

Table 2 Simulated effect of genotype and feeding environment on reproductive performance, lifespan and LTE relative to data from Langhill cows (mean ± standard deviation)

		Control			Select			HC			TC	
Trait	Observed	Simulated	<i>P</i> -value	Observed	Simulated	<i>P</i> -value	Observed	Simulated	<i>P</i> -value	Observed	Simulated	<i>P</i> -value
FSC	$45 \pm 7 (49)^1$	44±5	0.80	$39 \pm 5 (41)$	42 ± 5	0.37	$42 \pm 5 (49)$	43 ± 5	0.75	$41 \pm 7 (45)$	43±7	99.0
DFS	$72 \pm 14 (30)$	79±16	<0.05	$77 \pm 14 (30)$	84 ± 21	<0.05	$77 \pm 14 (37)$	81 ± 19	<0.05	$72 \pm 14 (26)$	81 ± 19	<0.05
DOP	$106 \pm 39 (40)$	$97 \pm 24$	<0.05	$124 \pm 39 (26)$	$101 \pm 26$	<0.05	$117 \pm 39 (29)$	$99 \pm 26$	<0.05	$113 \pm 39 (36)$	$99 \pm 26$	<0.05
⋾	$383 \pm 34 (44)$	$378 \pm 24$	0.14	$395 \pm 34 (32)$	$382 \pm 27$	<0.05	$389 \pm 34 (36)$	$380 \pm 26$	<0.05	$389 \pm 34 (36)$	$380 \pm 26$	<0.05
LSP	NA	$2856 \pm 1247$	NA	NA	$2225 \pm 889$	N	ΝΑ	$2538 \pm 2958$	Ą	NA	$2544 \pm 3137$	N
빔	Ν Α	45±5	ΑN	NA	48 ± 5	Ν Α	ΝΑ	$49 \pm 1.3$	A	NA	$47 \pm 1.5$	Ν

DFS = days from calving to first service (d); DOP = days from calving to conception or days open (d); CI = calving interval (d); LSP = lifespan (d); LTE = lifetime efficiency (%); HC = high Sample size of n = 5 for FSC in observed and simulated populations, n = 200 and 138 for remaining traits in simulated and observed populations, respectively. Values reported in the parentheses indicate the likelihood of the mean simulated value of that trait being part of the corresponding population of observed values (%) concentrate; LC = low concentrate; NA = not available. -SC = first service conception (%);

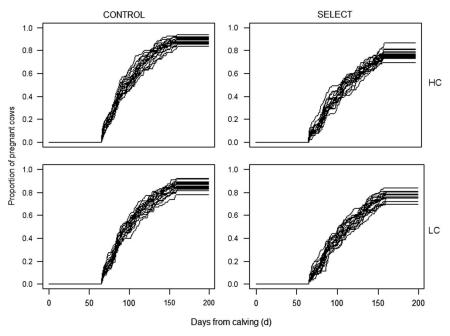


Figure 6 Simulated proportion of pregnant cows for the Langhill data set: HC = high concentrate and LC = low concentrate. Each line is one repetition of a simulation with a population of 200 cows.

more influence on LTE than the length of productive lifespan. Blank et al. (2012) found a strongly positive correlation between LTE (kg of milk/cow per day) and milk yield  $(R^2 = 0.8)$  of cows for the same lifespan, but no significant correlation between productive lifespan and LTE could be found ( $R^2$  not reported). The result in this study is also consistent with the spreadsheet model of Jones (2005) in which the financial impacts of reducing annual milk production to extend the productive life of dairy cows is calculated. The annual returns were estimated using specific assumptions on annual milk production, feed consumption, milk prices and feed prices. The simulated results showed that the net financial effect of sacrificing milk production for a longer productive life of dairy cows was generally a loss. However, neither the study by Jones (2005) nor the present study takes into account other economic benefits of having a longer productive life, such as more female calves available for replacement, fewer failed inseminations, or decreased veterinary costs due to fewer peri-parturient periods per year, nor do these models consider the societal concerns around animal welfare issues, which are especially important if production efficiency at the farm level is considered. VandeHaar (2014), for example, indicated that feed consumed by heifers can account for 15% to 30% of the amount of feed of a cow during her lifetime. The published estimates of the economic losses of clinical mastitis, for example, range from €61 to €97 per cow on a farm (Hogeveen et al., 2011). At the farm level, increased productive life and consequently longevity will lead to fewer replacement heifers being required to achieve the same level of output, thus reducing the replacement costs; more offspring are produced, and more intensive selection is possible (Bell et al., 2011). Garnsworthy (2004), using a modelling approach, reported

that improved longevity lowered the environmental impact; for example, 27% of methane produced on farms is expected to be produced by herd replacements, and thus by lowering the number of heifer replacements methane emissions would also be mitigated. Additionally, the relevance of including body composition to the calculation of feed efficiency is important when we want to compare two breeds differing largely in body composition. This, however, raises a nontrivial question of what value to put on body reserves. To our knowledge this has not satisfactorily been resolved, and is a complex issue as it most likely depends upon the time point at which those reserves are available; for example, body reserves at the start of lactation are presumably of more value for supporting milk production than in later lactation. As in this paper the major aim was to attempt to include the potential impacts of performance on the probability of reproducing, and thus variation of lifespan, such a guestion of body reserve value has not been addressed. These points of discussion indicate the need for further analysis of LTE not only in terms of the definition used at the individual level but also at the herd level if the consequences of future management strategies are fully to be evaluated.

#### **Conclusions**

The model used in this study consistently simulated the effects of productive performance on reproductive outcomes of various genotypes of cows across feeding systems. The simulated effect of genotypes on reproductive performance was in agreement with the observed trends in the test data. Genetic selection for greater milk production impaired the reproductive performance. The effect of a poorer nutritional environment on simulated reproductive performance was

relatively well modelled with respect to the test data. LTE, defined as the energy ratio between milk and feed intake, was slightly higher for high—genetic-merit cows compared with lower-genetic-merit cows due to the compensation in milk quantity over a longer productive life. However, this LTE definition does not include associated costs. The value of body reserves, economic indicators and other management factors should also be incorporated to fully and more accurately simulate lifetime profitability in different scenarios.

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#### References

Albarrán-Portillo B and Pollott GE 2013. The relationship between fertility and lactation characteristics in Holstein cows on United Kingdom commercial dairy farms. Journal of Dairy Science 96, 635–646.

Bell MJ, Wall E, Russell G, Simm G and Stott AW 2011. The effect of improving cow productivity, fertility, and longevity on the global warming potential of dairy systems. Journal of Dairy Science 94, 3662–3678.

Bergez JE, Chabrier P, Gary C, Jeuffroy MH, Makowski D, Quesnel G, Ramat E, Raynal H, Rousse N, Wallach D, Debaeke P, Durand P, Duru M, Dury J, Faverdin P, Gascuel-Odoux C and Garcia F 2013. An open platform to build, evaluate and simulate integrated models of farming and agro-ecosystems. Environmental Modelling and Software 39, 39–49.

Berry DP, Bermingham ML, Good M and More SJ 2011. Genetics of animal health and disease in cattle. Irish Veterinary Journal 64, 5.

Berry DP, Buckley F, Dillon P, Evans RD, Rath M and Veerkamp RF 2003. Genetic relationships among body condition score, body weight, milk yield, and fertility in dairy cows. Journal of Dairy Science 86, 2193–2204.

Blanc F and Agabriel J 2008. Modelling the reproductive efficiency in a beef cow herd: effect of calving date, bull exposure and body condition at calving on the calving—conception interval and calving distribution. The Journal of Agricultural Science 146. 143—161.

Blanc F, Martin GB and Bocquier F 2001. Modelling reproduction in farm animals: a review. Reproduction, Fertility and Development 13, 337–353.

Blank B, Schaub D, Paulsen HM and Rahmann G 2012. Herd parameters in organic and conventional dairy farms and their role in greenhouse gas emissions from milk production. Tackling the Future Challenges of Organic Animal Husbandry. Proceedings of the 2nd OAHC (ed. G Rahmman and D Godinho), pp. 269–272. Hamburg/Trenthorst, Germany.

Brun-Lafleur L, Cutullic E, Faverdin P, Delaby L and Disenhaus C 2013. An individual reproduction model sensitive to milk yield and body condition in Holstein dairy cows. Animal 7, 1332–1343.

Butler ST, Pelton SH and Butler WR 2006. Energy balance, metabolic status, and the first postpartum ovarian follicle wave in cows administered propylene glycol. Journal of Dairy Science 89, 2938–2951.

Butler WR 2000. Nutritional interactions with reproductive performance in dairy cattle. Animal Reproduction Science 60–61, 449–457.

Coffey MP, Simm G, Oldham JD, Hill WG and Brotherstone S 2004. Genotype and diet effects on energy balance in the first three lactations of dairy cows. Journal of Dairy Science 87, 4318–4326.

CRV 2014. Dutch cows excel in longevity. Retrieved December 4, 2014, from https://global.crv4all.com/aboutus/publications/highlights/79396/.

Cutullic E, Delaby L, Gallard Y and Disenhaus C 2011. Dairy cows' reproductive response to feeding level differs according to the reproductive stage and the breed. Animal 5, 731–740.

Darwash AO, Lamming GE and Woolliams JA 1997. Estimation of genetic variation in the interval from calving to postpartum ovulation of dairy cows. Journal of Dairy Science 80, 1227–1234.

Friggens NC, Chagunda MGG, Bjerring M, Ridder C, Hojsgaard S and Larsen T 2007. Estimating degree of mastitis from time-series measurements in milk: a test of a model based on lactate dehydrogenase measurements. Journal of Dairy Science 90, 5415–5427.

Friggens NC, Disenhaus C and Petit HV 2010. Nutritional sub-fertility in the dairy cow: towards improved reproductive management through a better biological understanding. Animal 4, 1197–1213.

Garnsworthy PC 2004. The environmental impact of fertility in dairy cows: a modelling approach to predict methane and ammonia emissions. Animal Feed Science and Technology 112, 211–223.

Hansen M, Lund MS, Pedersen J and Christensen LG 2004. Gestation length in Danish Holsteins has weak genetic associations with stillbirth, calving difficulty, and calf size. Livestock Production Science 91, 23–33.

Hare E, Norman HD and Wright JR 2006. Survival rates and productive herd life of dairy cattle in the United States. Journal of Dairy Science 89, 3713–3720.

Hogeveen H, Huijps K and Lam TJGM 2011. Economic aspects of mastitis: new developments. New Zealand Veterinary Journal 59, 16–23.

Horan B, Mee JF, O'Connor P, Rath M and Dillon P 2005. The effect of strain of Holstein-Friesian cow and feeding system on postpartum ovarian function, animal production and conception rate to first service. Theriogenology 63, 950–971.

Inchaisri C, Jorritsma R, Vos PL, van der Weijden GC and Hogeveen H 2010. Economic consequences of reproductive performance in dairy cattle. Theriogenology 74, 835–846.

Jones BL 2005. The economic trade-offs between annual milk production and the productive lives of dairy cows. University of Wisconsin, Centre for Dairy Profitability.

Knaus W 2009. Dairy cows trapped between performance demands and adaptability. Journal of the Science of Food and Agriculture 89, 1107–1114.

Kuhn MT, Hutchison JL and Wiggans GR 2006. Characterization of Holstein heifer fertility in the United States. Journal of Dairy Science 89, 4907–4920.

Leroy JLMR, Van Soom A, Opsomer G, Goovaerts IGF and Bols PEJ 2008. Reduced fertility in high-yielding dairy cows: are the oocyte and embryo in danger? Part II. Mechanisms linking nutrition and reduced oocyte and embryo quality in high-yielding dairy cows. Reproduction in Domestic Animals 43, 623–632.

Löf E, Gustafsson H and Emanuelson U 2012. Evaluation of two dairy herd reproductive performance indicators that are adjusted for voluntary waiting period. Acta Veterinaria Scandinavica 54, 5.

Lormore MJ and Galligan DT 2001. Economics of atypical lactation. Journal of Dairy Science 84, E212–E215.

Martin O and Sauvant D 2010. A teleonomic model describing performance (body, milk and intake) during growth and over repeated reproductive cycles throughout the lifespan of dairy cattle. 1. Trajectories of life function priorities and genetic scaling. Animal 4, 2048–2056.

McGowan MR, Veerkamp RF and Anderson L 1996. Effects of genotype and feeding system on the reproductive performance of dairy cattle. Livestock Production Science 46, 33–40.

McNeill RE, Diskin MG, Seenan JM and Morris DG 2006. Associations between milk progesterone concentration on different days and with embryo survival during the early luteal phase in dairy cows. Theriogenology 65, 1435–1441.

Oltenacu PA and Algers B 2005. Selection for increased production and the welfare of dairy cows: are new breeding goals needed? Ambio 34, 4–5.

Pedernera M, García SC, Horagadoga A, Barchia I and Fulkerson WJ 2008. Energy balance and reproduction on dairy cows fed to achieve low or high milk production on a pasture-based system. Journal of Dairy Science 91, 3896–3907.

Phuong HN, Martin O, de Boer IJM, Ingvartsen KL, Schmidely P and Friggens NC 2015. Deriving estimates of individual variability in genetic potentials of performance traits for 3 dairy breeds, using a model of lifetime nutrient partitioning. Journal of Dairy Science 98, 618–632.

Pritchard T, Coffey MP, Mrode R and Wall E 2013. Understanding the genetics of survival in dairy cows. Journal of Dairy Science 96, 3296–3309.

## Phuong, Blavy, Martin, Schmidely and Friggens

Pryce JE, Coffey MP and Simm G 2001. The relationship between body condition score and reproductive performance. Journal of Dairy Science 84, 1508–1515.

Pryce JE, Nielsen BL, Veerkamp RF and Simm G 1999. Genotype and feeding system effects and interactions for health and fertility traits in dairy cattle. Livestock Production Science 57, 193–201.

R Development Core Team 2015. The GNU Project. The R Project for Statistical Computing. Retrieved May 4, 2015, from http://www.rproject.org/

Seegers H, Beaudeau F, Fourichon C and Bareille N 1998. Reasons for culling in French Holstein cows. Preventive Veterinary Medicine 36, 257–271.

Senatore EM, Butler WR and Oltenacu PA 2010. Relationships between energy balance and post-partum ovarian activity and fertility in first lactation dairy cows. Animal Science 62, 17–23.

Spicer LJ 2001. Leptin: a possible metabolic signal affecting reproduction. Domestic Animal Endocrinology 21, 251–270.

Spicer LJ, Alpizar E and Echternkamp SE 1993. Effects of insulin, insulin-like growth factor I, and gonadotropins on bovine granulosa cell proliferation, progesterone production, estradiol production, and (or) insulin-like growth factor I production in vitro. Journal of Animal Science 71, 1232–1241.

VandeHaar MJ 2014. Feeding and breeding for a more efficient cow. WCDS Advances in Dairy Technology 26, 17–30.

VandeHaar MJ and St-Pierre N 2006. Major advances in nutrition: relevance to the sustainability of the dairy industry. Journal of Dairy Science 89, 1280–1291.

Wall E 2008. Genetic and environmental effects on fitness traits in dairy cattle. EAAP Meeting, Vilnius. Retrieved November 15, 2014, from http://www.eaap.org/Previous\_Annual\_Meetings/2008Vilnius/Papers/published/26\_Wall.pdf

Walsh S, Buckley F, Pierce K, Byrne N, Patton J and Dillon P 2008. Effects of breed and feeding system on milk production, body weight, body condition score, reproductive performance, and postpartum ovarian function. Journal of Dairy Science 91, 4401–4413.

Wiltbank M, Lopez H, Sartori R, Sangsritavong S and Gümen A 2006. Changes in reproductive physiology of lactating dairy cows due to elevated steroid metabolism. Theriogenology 65, 17–29.

Wright IA, Rhind SM, Whyte TK and Smith AJ 1992. Effects of body condition at calving and feeding level after calving on LH profiles and the duration of the post-partum anoestrous period in beef cows. Animal Science 55, 41–46.

Yániz J, López-Gatius F, Bech-Sàbat G, García-Ispierto I, Serrano B and Santolaria P 2008. Relationships between milk production, ovarian function and fertility in high-producing dairy herds in north-eastern Spain. Reproduction in Domestic Animals 43. 38–43.