ABSTRACT

Body condition score (BCS) records of primiparous Holstein cows were analyzed both as a single measure per animal and as repeated measures per sire of cow. The former resulted in a single, average, genetic evaluation for each sire, and the latter resulted in separate genetic evaluations per day of lactation. Repeated measure analysis yielded genetic correlations of less than unity between days of lactation, suggesting that BCS may not be the same trait across lactation. Differences between daily genetic evaluations on d 10 or 30 and subsequent daily evaluations were used to assess BCS change at different stages of lactation. Genetic evaluations for BCS level or change were used to estimate genetic correlations between BCS measures and fertility traits in order to assess the capacity of BCS to predict fertility. Genetic correlation estimates with calving interval and non-return rate were consistently higher for daily BCS than single measure BCS evaluations, but results were not always statistically different. Genetic correlations between BCS change and fertility traits were not significantly different from zero. The product of the accuracy of BCS evaluations with their genetic correlation with the UK fertility index, comprising calving interval and non-return rate, was consistently higher for daily than for single BCS evaluations, by 28 to 53%. This product is associated with the conceptual correlated response in fertility from BCS selection and was highest for early (d 10 to 75) evaluations. (Key words: body condition score, body condition change, random regression, fertility index)

INTRODUCTION

Long-term selection emphasis on milk yield combined with its antagonistic genetic correlation with female fertility (Pryce et al., 1998; Pryce and Veerkamp, 2001; Roxström et al., 2001) has resulted in a reduced capacity of dairy cows to conceive and carry a calf to term (Royal et al., 2000). Incorporating genetic evaluations for fertility traits into breeding programs (Pryce and Veerkamp, 2001) could help alleviate the problem.

The accuracy of genetic evaluations depends on the number of validated records and the heritability of a trait. In the UK, only about 10% of milk-recorded herds maintain accurate insemination data (Kadarmideen and Coffey, 2001). Further, the heritability of most fertility traits is quite low, typically <0.05 (Wall et al., 2003). Hence, in an environment where the recording of fertility traits is uncertain, indirect measures of fertility must be considered to increase the accuracy of genetic evaluations. Such information may be found in other traits with higher heritability, more readily available records, and considerable genetic correlations with fertility. Milk yield (Harrison et al., 1990; Pryce and Veerkamp, 2001), linear type traits (Pryce et al., 2000), and BCS (Pryce et al., 2000; Dechow et al., 2001; Berry et al., 2003) meet these criteria. Analyzing fertility jointly with milk yield and BCS may also account, to a certain extent, for selection and management bias. Wall et al. (2003) jointly analyzed 4 fertility traits, milk yield, and BCS and produced preliminary fertility evaluations of bulls used in the UK.

Body condition score changes throughout the lactation of a cow, responding to changes in her energy balance (Coffey et al., 2003). As milk yield peaks and demand for energy exceeds intake of energy, the cow mobilizes her lipid reserves and gets thinner, thereby compromising her body condition. This process is related to the daily milk yield curve, which is almost...
exactly opposite to the energy balance and BCS curves (Coffey et al., 2002, 2003). Excessive mobilization of body reserves may have adverse effects on cow fertility and health (Collard et al., 2000; de Vries and Veerkamp, 2000). In fact, inseminations in the UK usually start when daily milk yield approaches its peak, while energy balance and BCS are on the decline. This pattern suggests that BCS at about the time cows are being inseminated might be a more informative fertility predictor than BCS at other stages of lactation. Also, the rate that a cow loses body energy in the beginning of the lactation might be related to her propensity to conceive, although results of Pryce et al. (2001) do not fully support this assertion.

In the UK, field officers of Holstein UK record BCS once for each cow, during their type classification visits to participating herds. First-lactation cows are scored on a scale of 1 to 9, where 1 = thin and 9 = fat. This score is adjusted for recording officer by scaling records so that individual field officer standard deviations are equal to the mean standard deviation of all field officers (Brotherstone, 1994; Jones et al., 1999). Cows are scored only once, each at a different stage of lactation. This single measurement does not capture the BCS variation throughout the cow’s lactation. Furthermore, BCS changes cannot be monitored at the cow level.

The objective of this study was to evaluate, using appropriate models, single lactation and daily measures of BCS level and BCS change throughout lactation as fertility indicator traits.

MATERIALS AND METHODS

Data Description

A data set with cow BCS records was obtained from Holstein UK. Edits removed records taken after d 305 of lactation, and cows that had their first calving outside the 18- to 40-mo age range. Further edits removed records in herd-year-seasons of visit (HYS) with size < 2 and records of cows whose sires had < 2 daughters. Three seasons of visit were defined as follows: January to April, May to August, and September to December (inclusive). The final data set consisted of 206,613 BCS cow records from 3934 sires distributed in 22,703 HYS. All cows were in their first lactation and had their first calving from 1996 through 2002. Figure 1 illustrates the distribution of records and BCS across the 305 d of lactation. On average, cows were visited and classified 132 d after they had calved.

The BCS data set was matched to datasets from National Milk Records and Cattle Information Service containing milk yield and fertility data. Fertility traits included calving interval (CI) between first and second calving and non-return rate (NR). The latter was recorded as 0 if the cow was re-inseminated 56 d after the first insemination and as 1 if the cow did not return to service in that period. Wall et al. (2003) described in detail the validation rules for insemination traits in the UK. These 2 fertility traits form the UK fertility index. Milk yield was daily yield at test nearest to d 110 of lactation (DMY), which was used in the preliminary genetic evaluation for fertility (Wall et al., 2003). All cows had BCS and DMY records, whereas 73 and 92%, respectively, had CI and NR records. Table 1 summarizes the data used in this study.

BCS Data Analyses

Body condition score was analyzed as repeated measures per sire. This was possible given the distribution of sire daughter BCS throughout the lactation. Because there was only one observation per cow, a sire rather than animal model was used. Data were analyzed with the following random regression model:

\[
y_{hijkl} = MC_k + b_p \cdot p_{ch} + b_{1s} \cdot AC + b_{2a} \\
\cdot AC^2 + m_{q1} + b_m \cdot DMY + b_d \cdot \text{days}
\]

where \(y_{hijkl} = \text{BCS of daughter } h \text{ of sire } i \text{ on day } j \text{ of lactation in management class } k\); \(MC_k = \text{fixed effect of management class } k\) defined as either HYS or the HYS

<table>
<thead>
<tr>
<th>Trait (unit)</th>
<th>Records (no.)</th>
<th>Mean</th>
<th>SD</th>
</tr>
</thead>
<tbody>
<tr>
<td>BCS (score)</td>
<td>206,613</td>
<td>4.47</td>
<td>1.57</td>
</tr>
<tr>
<td>CI (d)</td>
<td>150,301</td>
<td>396.39</td>
<td>55.22</td>
</tr>
<tr>
<td>NR (0/1)</td>
<td>190,407</td>
<td>0.67</td>
<td>0.47</td>
</tr>
<tr>
<td>DMY (kg)</td>
<td>206,613</td>
<td>25.77</td>
<td>5.81</td>
</tr>
</tbody>
</table>
by day of lactation interaction (HYSd); \(b_p\) = fixed linear regression on percentage North American Holstein genes (pch); \(b_{1a}\) and \(b_{2a}\) = fixed linear and quadratic regression on age at calving (AC), respectively; \(mo\) = fixed effect of month of calving \(l\); \(b_m\) and \(b_d\) = fixed linear regressions on DMY and corresponding days of lactation (days), respectively; \(d_{jm}\) = Legendre polynomial \(m\) of BCS on day \(j\) of lactation, \(\alpha_{jm}\) = random regression coefficient on sire \(i\) with \(\sigma^2_{ij}\); \(\beta_{jm}\) = fixed regression coefficient of BCS on day \(j\) of lactation, \(\sigma^2_{ij}\) = random regression coefficient on sire \(i\) with \(\sigma^2_{ij}\); \(A\cdot G\) (A = sire relationship matrix); and \(e_{ijkl}\) = random residual with \(\sigma^2_{ij}\). Based on a previous study with similar data (Coffey et al., 2003), the order of the fixed and random regression polynomials was set to 5 (quartic including intercept) and 4 (cubic including intercept), respectively. Based on the study of Coffey et al. (2003), 10 measurement error classes were defined as follows: \(d\) 1 to 15, 16 to 29, 30 to 59, 60 to 89, 90 to 119, 120 to 149, 150 to 179, 180 to 209, 210 to 239, and 240 to 305 of lactation. Different residual variances were estimated for each class. Residual co-variances were assumed to be zero. Variance components were estimated with the ASYREML software package (Gilmour et al., 1998).

Sire variances of BCS on day \(j\) of lactation \((V_j)\) were estimated as \(V_j = d_j' \cdot G \cdot d_j\), where \(d_j\) is the polynomial vector for day \(j\) and \(G\) is as defined in model [1]. Similarly, sire BCS co-variances between days \(i\) and \(j\) \((COV_{ij})\) were estimated as \(COV_{ij} = d_i' \cdot G \cdot d_j\). Sire variance of BCS change from day \(i\) to \(j\) was then estimated as the sum of individual day variances minus twice their co-variance.

Prediction error variances per sire for day \(j\) of lactation \((PEV_{ij})\) were estimated as \(PEV_{ij} = d_i' \cdot S \cdot d_j\), where \(d_i\) is the polynomial vector for day \(i\) and \(S\) is the square of the diagonal standard error matrix for each sire solution. Sire reliability for day \(j\) was then calculated as \(1 - (PEV_{ij}/V_j)\). Similarly, prediction error co-variances per sire between days \(i\) and \(j\) \((PECOV_{ij})\) were estimated as \(PECOV_{ij} = d_i' \cdot S \cdot d_j\). Prediction error variance of BCS change from day \(i\) to \(j\) was then estimated as the sum of individual day \(PEV\) minus twice \(PECOV\), and reliability of BCS change was calculated from this and the corresponding sire variance.

Body condition score was also analyzed as a single lactation measure with single-trait model [2]:

\[
y_{ijkl} = MC_k + b_p \cdot pch + b_{1a} \cdot AC + b_{2a} \\
\cdot AC^2 + mo + b_m \cdot DMY + b_d \cdot days + b_{c} \\
\cdot dcs + sire + e_{ijkl}
\]

where \(MC_k\) = fixed effect of HYS k, \(b_{c}\) = fixed linear regression on days of lactation when the cow was scored for body condition (dcs), and \(sire\) = random effect of sire \(i\). All other effects were as defined in model [1].

**Correlation Between BCS and Fertility**

Sire evaluations for each day of lactation calculated with model [1] were used to estimate genetic correlations between daily BCS and fertility with bivariate analyses based on model [3]:

\[
y_1_{hikl}, y_2_{hikl} = MC_k + b_p \cdot pch + b_{1a} \cdot AC \\
+ b_{2a} \cdot AC^2 + mo + b_m \cdot DMY + b_B \cdot dBCS + e_{hikl}
\]

where \(y_1 = DMY\) and \(y_2 = CI\) or NR record of daughter of sire \(i\), \(MC_k\) = fixed effect of herd-year-season of calving \(k\), and \(b_B\) = fixed linear regression on the daily sire genetic evaluation for BCS from model [1] weighted by its reliability (dBCS). Other effects were as in model [1]. The analysis was repeated 305 times, once for each day of lactation. Although we were interested in the genetic correlation between BCS and fertility traits, DMY was included to use all records available and account for selection bias. The genetic correlation between daily BCS and fertility traits was calculated by multiplying the estimated regression on daily BCS evaluation from model [3] with the ratio of genetic standard deviations of daily BCS over CI or NR. We chose this procedure because we wanted to estimate the genetic correlation between a trait analyzed as repeated measures (BCS) and traits analyzed as single measures (CI and NR). We were not convinced that our software would accommodate a multiple-trait (BCS and fertility) analysis in this context. Brotherstone and Hill (1991) and Pryce et al. (2000) followed a similar procedure to estimate correlations between herd life and type traits, and BCS and CI, respectively. The procedure may be sensitive to genetic variances assumed. In this case, genetic variances of daily BCS were those that had been estimated with model [1]. Genetic variances of fertility traits were estimated with model [4] shown subsequently. Standard errors of the genetic correlation estimates were calculated as the product of the point estimate times the square root of the sum of the ratios of the variance of the regression coefficient estimate over the estimate squared and the variances of the 2 genetic variance estimates over 4 times the respective estimates squared. This formula assumes no co-variances between the regression coefficient and the genetic variance estimates.
where $y_1 = \text{BCS}$, $y_2 = \text{CI or NR record of daughter h of sire i}$, and $MC_k = \text{fixed effect of herd-year-season of visit k (for BCS) or calving (for CI and NR)}$. All other effects were as in model [2]. Genetic correlations between single measures of BCS and fertility traits were estimated with model [4].

The genetic correlation between fertility and BCS change from the beginning of the lactation was estimated by sequentially fitting in model [3] the difference of sire genetic evaluations on d 11 to 305 from the evaluation on d 10, weighted by the corresponding reliability. Day 10 was considered as the reference day representing the beginning of lactation because earlier days were largely missing any validated BCS records. The exercise was repeated using sire genetic evaluation difference between d 30 and d 31 through 305.

### RESULTS

#### BCS Data Analyses

Fixed regression estimates of BCS on days of lactation are shown in Figure 2. For both management group models, BCS decreased during the first 2 to 3 mo of lactation and increased afterward. When the model included a HYS management group effect, BCS increase was greater than the initial drop. When HYSd defined the management group, BCS increased to lower levels than its initial value at the beginning of lactation. The latter is considered a more accurate description of the BCS curve. When the management group includes the stage of lactation (as HYSd does) it accounts for the physiological BCS change during the lactation (Figure 1). Otherwise, cows at a low energy stage (e.g., d 40 to 100 in lactation) would be unfavorably compared with their herdmates who are toward the end of their lactation and in the same contemporary group. Fitting a fixed regression on days in lactation only partially accounts for this. In addition, when the model includes HYS, fixed curves are allowed to vary among management groups by a constant. When a HYSd is fitted instead, fixed curves are also allowed to vary in slope, which goes some way to allowing all coefficients of the fixed regression in model [1] to vary depending on management group. Heritability of BCS estimated with models [1] and [2] is given in Figure 3. Single measure BCS heritability was tangential to the heritability curve estimated with random regression models. Daily heritability estimates were higher at either end of the curve as result of higher sire variance estimates in early and, less so, late lactation. Table 2 shows estimates of sire variance for selected days of lactation. Veerkamp et al. (2001) also reported changing genetic variance estimates throughout lactation from random regression model analysis. In their study, however, early lactation (d 1 to 50) genetic variance estimates were not higher than estimates in the remaining lactation. Small numbers of records at either end of the lactation period (Figure 1) may compromise the accuracy of these estimates. Nevertheless, in our study, daily sire variance estimates at various stages of lactation differed significantly from each other (Table 2).

Average sire reliabilities per day of lactation estimated with model [1] ranged from 0.27 in the beginning (d 1 to 50) and end (d > 250) of the lactation to 0.48 in the middle (d 100 to 180) of the lactation. Higher estimates in midlactation were consistent with a heav-
Table 2. Sire variance estimates for daily BCS (selected days) from random regression analysis including a herd-year-season (HYS) or herd-year-season by days of lactation interaction (HYSd) management group and corresponding regressions of calving interval (CI) and non-return rate (NR) on daily BCS.

<table>
<thead>
<tr>
<th>Day of lactation</th>
<th>Variance estimate$^1$</th>
<th>CI Regression$^2$</th>
<th>NR Regression$^3$</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>HYS</td>
<td>HYSd</td>
<td>HYS</td>
</tr>
<tr>
<td>10</td>
<td>0.132</td>
<td>0.195</td>
<td>-4.19</td>
</tr>
<tr>
<td>70</td>
<td>0.103</td>
<td>0.114</td>
<td>-4.79</td>
</tr>
<tr>
<td>150</td>
<td>0.117</td>
<td>0.122</td>
<td>-4.27</td>
</tr>
<tr>
<td>280</td>
<td>0.143</td>
<td>0.139</td>
<td>-3.64</td>
</tr>
</tbody>
</table>

$^1$SE = 0.006 to 0.018.
$^2$SE = 0.40 to 0.80; CI sire variance, 20.0 ± 2.8.
$^3$SE = 0.005 to 0.007; NR sire variance, 0.00044 ± 0.0001.

Correlation Between BCS and Fertility

Genetic correlations among daily BCS records were near unity for adjacent days or days close to each other and decreased as the distance between days increased. Genetic correlations of BCS in early lactation (d 1 to 50) were 0.70 to 0.75 with midlactation (d 100 to 180) BCS, and were 0.50 to 0.60 with late (d > 250) lactation BCS. Genetic correlations between mid and late lactation BCS ranged from 0.81 to 0.96. These results suggest that the trait changes genetically as lactation progresses.

Correlation Between BCS and Fertility

Genetic correlations between daily BCS and CI estimated with model [3] and between single measure BCS and CI from model [4] are shown in Figure 4. Genetic correlations ranged from −0.35 to −0.31 (SE = 0.06 to 0.07) with the HYS model and from −0.37 to −0.34 (SE = 0.06 to 0.07) with the HYSd model. Correlations were strongest at around d 50 to 70 of the lactation and weakest toward the end, consistent with the size of the corresponding regression coefficients (see Table 2 for selected estimates). However, the standard errors of these estimates suggest that differences were not significant. Correlation estimates of daily BCS were slightly higher than the single measure BCS estimate of −0.31 (SE = 0.07), but differences were not significant.

Genetic correlations between change of daily BCS from d 10 and CI estimated with model [3] (not shown) ranged from 0.07 to 0.14 (SE = 0.08 to 0.09) with the HYS model and from 0.05 to 0.08 (SE = 0.07 to 0.08) with the HYSd model. Estimates slightly increased with progressing lactation, but their large standard errors render them non-significantly different from zero. When daily BCS change from d 30 was considered instead of d 10, results (not shown) were practically the same.

Genetic correlations between daily BCS and NR estimated with model [3] and between single measure BCS and NR from model [4] are shown in Figure 5. Genetic correlations ranged from −0.32 to −0.19 (SE = 0.08 to 0.11).
times its genetic correlation with fertility from model 4. In this respect, daily BCS measures from random regression model analysis either with HYS or HYSd constantly out-performed single measure BCS by 28 to 53%.

**DISCUSSION**

A constant body condition score is associated with the ability of the cow to produce milk while maintaining its energy balance. Because of changes in daily milk yield throughout the lactation and other physiological changes, including reproduction and pregnancy status, the energy balance profile varies, prompting variation of daily BCS (Coffey et al., 2003). In fact, Coffey et al. (2003) found that the lowest body energy level occurs, on average, 3 mo postpartum, very close to the peak of daily milk yield. In the present study, the fixed regression of BCS on days of lactation was lowest on d 72 and 87 for HYS and HYSd models, respectively, reflecting the lowest body energy levels. Because the HYSd model allows for better comparisons both within and between management groups, the value of 87 is accepted as the more correct of the 2, suggesting that the lowest adjusted phenotypic BCS is expected to occur on this day, very close to the energy balance nadir suggested by Coffey et al. (2003).

The changing genetic profile of BCS throughout the lactation, observed in this study, was illustrated by the heritability curve and genetic correlation estimates below unity among different daily BCS. Results were, in general, consistent with previous studies (Jones et al., 1999; Berry et al., 2002) and suggest that BCS is not genetically the same trait in various stages of lactation. This is not surprising given the BCS association with energy balance. The daily BCS variation was harnessed using random regression models, and distinct genetic parameters were estimated for each daily measure.

The relationship of energy balance with reproduction and fertility has been documented in the literature (Collard et al., 2000; de Vries and Veerkamp, 2000; Veerkamp et al., 2000). Negative energy balance is associated with difficulties for the cow to conceive and maintain the calf. Unfortunately, body energy is at its lowest when inseminations have usually just started, often leading to fertility problems particularly in high-yielding cows. Therefore, BCS measured at that time might have been expected to give better than average information on cow fertility. Results of this study partially corroborate this claim. Genetic correlations between BCS and either CI or NR were consistently higher for daily than for single lactation measure of BCS. Differences, however, were mostly nonsignificant. Genetic correla-
tions were strongest at the early stages of lactation, suggesting that early daily BCS measures are probably better predictors of fertility. This is related to the animal genetic component being more pronounced at early stages of lactation, as suggested by higher heritability and sire variance estimates from a random regression model. Strong regressions of cow fertility on genetic merit of sires for BCS in early lactation further support the claim. These results are in partial agreement, at least regarding the relative strength of the correlation estimate, with Veerkamp et al. (2001) and Berry et al. (2003). In our study, both HYS and HYSd models yielded similar estimates, suggesting that the definition of management group is not expected to affect the predictive power of daily BCS.

Looking for the best genetic indicator of a trait is conceptually equivalent to selecting for the predictor to elicit a correlated response in the trait of interest. In this case, BCS is the predictor, and fertility is the trait we wish to improve genetically. The correlated response then becomes a function of the genetic correlation between BCS and fertility and the accuracy (square root of the reliability) of the BCS genetic evaluation. Fertility in the UK is considered as the combination of CI and NR, with approximate weights of 1:5 on the 2 traits. These weights were placed on the product of genetic correlation between BCS and CI or NR and accuracy of BCS evaluation and compared for the various BCS measures. When BCS was analyzed as a repeated measure with random regressions, its predictive capacity was increased by 28 to 53% compared with single measure BCS analysis. The improvement was more pronounced at early stages of lactation and stabilized after d 75. The observed superiority of the random regression over single measure analysis results is attributed to NR, which is the dominant trait in the fertility index. Analyzing BCS with random regression models, compared with single measure analysis, improved the genetic correlation with NR more than with CI.

The change in BCS evaluation across lactation did not have high genetic correlation with fertility traits. In all cases, genetic correlation estimates were considerably lower than estimates for BCS level, supporting earlier reports by Pryce et al. (2001) and Berry et al. (2003). A possible explanation may be that each cow has a genetically predetermined lowest level of body energy (and BCS) that she is allowed to reach, and it is this nadir that determines her aptitude for fertility. The speed of reaching this level seems to be less important than the level itself. Results of this study suggest that BCS change across lactation is not as useful as BCS level for predicting fertility.

**CONCLUSIONS**

Based on the results of this study, BCS analyzed either as repeated or single measures can be used as a predictor of fertility. Correlations between fertility and BCS changes across lactation were not significantly different from zero. Daily BCS parameters derived from random regression analyses suggested that measures obtained in the first stage (about d 75) of lactation are the best predictors. This is primarily because of the stronger genetic correlation with NR and the relatively large weight the current UK fertility index places on this trait. Analysis of field-recorded BCS as repeated measures per sire using the HYSd model would be preferable because it captures the variation of the trait across lactation. Improving the consistency of BCS recording in the first 3 mo of lactation would facilitate its use as a predictor of fertility.

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