Heritability of rectal temperature and genetic correlations with production and reproduction traits in dairy cattle

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ABSTRACT

Genetic selection for body temperature during heat stress might be a useful approach to reduce the magnitude of heat stress effects on production and reproduction. Objectives of the study were to estimate the genetic parameters of rectal temperature (RT) in dairy cows in freestall barns under heat stress conditions and to determine the genetic and phenotypic correlations of rectal temperature with other traits. Afternoon RT were measured in a total of 1,695 lactating Holstein cows sired by 509 bulls during the summer in North Florida. Genetic parameters were estimated with Gibbs sampling, and best linear unbiased predictions of breeding values were predicted using an animal model. The heritability of RT was estimated to be 0.17 ± 0.13. Predicted transmitting abilities for rectal temperature changed 0.0068 ± 0.0020°C/yr from (birth year) 2002 to 2008. Approximate genetic correlations between RT and 305-d milk, fat, and protein yields, productive life, and net merit were significant and positive, whereas approximate genetic correlations between RT and somatic cell count score and daughter pregnancy rate were significant and negative. Rectal temperature during heat stress has moderate heritability, but genetic correlations with economically important traits mean that selection for RT could lead to lower productivity unless methods are used to identify genes affecting RT that do not adversely affect other traits of economic importance.

Key words: heritability, rectal temperature, heat stress

INTRODUCTION

Heat stress has adverse effects on milk production and reproduction of dairy cattle (Kadzere et al., 2002; West, 2003; Hansen, 2007). As noted by Hansen (2007), the problem of heat stress is growing because of increases in milk yield that result in higher metabolic heat production and because of anticipated changes in the global climate. Genetic selection for core body temperature regulation is one potential strategy to mitigate effects of heat stress on dairy cows. Most of the negative effects of heat stress on animal performance are either a consequence of the physiological adaptations that homeotherms undergo to regulate body temperature or the adverse consequences of failure to regulate body temperature (Hansen, 2011). Thus, an animal with genetic adaptations that make for more effective regulation of core body temperature will experience a less severe reduction in production. This idea has been documented in dairy cattle by comparing milk production in a hot climate for cows that are genetically superior for body temperature regulation by virtue of inheritance of the slick hair gene (Olson et al., 2003) or a specific allele of ATP1A1 (Liu et al., 2011). In both cases, cattle with the genotype that favors body temperature regulation had greater milk yield than cattle with the less favorable genotype.

Core body temperature is commonly ascertained by measuring rectal temperature. If rectal temperature (RT) during heat stress is heritable, it might be possible to improve thermostolerance of dairy cattle by selection for this trait. Heritability of RT in dairy cattle was reported to range from 0.15 to 0.31 (Seath, 1947), but changes in production level and environment since that study was conducted may have caused a change in heritability.

The current study had 2 objectives. The first was to estimate the genetic parameters of RT in dairy cattle subjected to heat stress. The second was to estimate genetic and phenotypic correlations of RT with production and reproduction traits.

MATERIALS AND METHODS

Rectal Temperature

The study was conducted with lactating Holstein cows at 3 dairies: the University of Florida Dairy Unit...
Table 1. Average (±SE) and range of animal and environmental variables used in the analysis

<table>
<thead>
<tr>
<th>Item</th>
<th>Mean (±SE)</th>
<th>Range</th>
</tr>
</thead>
<tbody>
<tr>
<td>Rectal temperatur (°C)</td>
<td>39.0 ± 0.6</td>
<td>37.2 to 41.7</td>
</tr>
<tr>
<td>Daily milk yield (kg)</td>
<td>32.5 ± 12.5</td>
<td>3.0 to 80.0</td>
</tr>
<tr>
<td>Parity</td>
<td>2.1 ± 0.9</td>
<td>1 to 4</td>
</tr>
<tr>
<td>DIM</td>
<td>172.2 ± 126.4</td>
<td>3 to 896</td>
</tr>
<tr>
<td>Tdb (°C)</td>
<td>30.6 ± 3.4</td>
<td>18.5 to 38.9</td>
</tr>
<tr>
<td>RH (%)</td>
<td>63.2 ± 11.5</td>
<td>32.5 to 92.4</td>
</tr>
<tr>
<td>Tdp (°C)</td>
<td>22.5 ± 2.9</td>
<td>9.1 to 28.1</td>
</tr>
<tr>
<td>Tbg (°C)</td>
<td>30.8 ± 3.6</td>
<td>18.5 to 38.6</td>
</tr>
<tr>
<td>THI</td>
<td>81.0 ± 4.3</td>
<td>65.1 to 88.2</td>
</tr>
</tbody>
</table>

THI = (1.8 × Tdb + 32) – [(0.55 - 0.0055 × RH) × (1.8 × Tdb - 26.8)].

Performance and Pedigree Data

Rectal temperatures were matched with performance data from the test-day closest to the date on which RT measurements were taken and included production and reproduction data for first through fifth lactations collected through the national milk recording program. Data were taken from the national dairy database (NDDB) maintained by the Animal Improvement Programs Laboratory (Beltsville, MD) and included PTA of milk, fat, and protein yields; SCS; productive life (PL; VanRaden and Wiggans, 1995); daughter pregnancy rate (DPR; VanRaden et al., 2004); and lifetime net merit (NM$; Cole et al., 2010). A pedigree file, including male and female ancestors back to 1960, was obtained from the NDDB and included 3,381 individuals. Summary statistics for daily milk yield on the days when RT was measured, parity, DIM, and environmental variables are shown in Table 1.

Genetic (Co)variances and PTA for RT

Mixed model analysis was performed using PROC MIXED of SAS (version 9.1.3; SAS Institute Inc., Cary, NC) with RT as the dependent variable to identify factors that should be included in the genetic model. Explanatory variables in the initial model included year of data collection, farm, parity, stage of lactation [separated into 3 stage of lactation classes: DIM < 100 (n = 741); DIM between 100 and 200 (n = 389); and DIM > 200 (n = 565)], and THI. The final model included effects of parity (P < 0.0001), stage of lactation (P < 0.0001), and farm (P < 0.0001) as class variables, and THI (P < 0.0001) and test-day milk yield (P < 0.10) as covariates.

The model used for estimation of genetic and environmental effects was

\[ y = Xb + Za + e, \]

where \(y\) is a vector of RT, \(b\) is a vector of fixed effects, \(a\) is a vector of additive genetic effects, and \(e\) is a vector of random residual effects. Fixed effects included parity, year of data collection, stage of lactation, and farm; THI and milk yield at the time of measurement were included as covariates. \(X\) and \(Z\) are incidence matrices relating a record to fixed environmental effects in \(b\) and to a random animal effect in \(a\), respectively. (Co)variances for RT were estimated using the Gibbs sampling method with the GIBBS1F90 software package (Misztal et al., 2002). Prior distributions were flat for the fixed effects and normal for the animal effect.
One Gibbs chain of 100,000 samples was drawn, the first 10,000 samples were discarded as burn-in, and then every tenth sample from the remaining 90,000 samples was used to calculate posterior means of the (co)variance components. Posterior means and standard deviations of the (co)variances were computed using POSTGIBBSF90 (Misztal et al., 2002). Plots of the Gibbs samples for each of the random effects were examined and no trends were observed, and it was concluded that a longer burn-in period was not needed. The heritability of RT was calculated from the resulting (co)variances, and its standard error was calculated using the Delta method (Lynch and Walsh, 1998).

Predicted transmitting abilities and reliabilities for RT were obtained using the MTDFREML software package (Boldman et al., 1993), which provides BLUP of breeding values upon convergence. (Co)variance components were those calculated in the previous step. Convergence was defined as a change in −2(log-likelihood) between successive rounds of iteration of 10⁻⁶ or less, and was achieved after 10⁸ rounds of iteration. Summary statistics for daily milk yield on the days when RT was measured, parity, DIM, and environmental variables are shown in Table 1.

**Phenotypic and Genetic Correlations**

Data on individual cows were used to calculate Pearson correlations for phenotypic values of RT, yield, SCS, DPR, and PL (RT measured on the farm and other values obtained from NDDB).

Data on RT were available for only 1,695 phenotypes, so genetic correlations of RT with the other traits were approximated using the method of Calo et al. (1973), rather than by a series of pairwise models. Genetic correlations were calculated from the PTA for 148 bulls included in the pedigree file that had a reliability for NM$ of at least 0.85 (taken from the NDDB) and a reliability for RT at least 0.25 (average reliability of RT = 0.289 ± 0.04). Genetic correlations were estimated as a function of correlations among PTA and their reliabilities (n = 148 bulls) using the method of Calo et al. (1973):

\[
\hat{r}_{g1,2} = \sqrt{\frac{(\sum RL_1 \times \sum RL_2)}{\sum (RL_1 \times RL_2)}} \times r_{1,2},
\]

where \(\hat{r}_{g1,2}\) = approximate genetic correlation between traits 1 and 2; \(\sum RL_1\) and \(\sum RL_2\) = the sum of reliabilities of traits 1 and 2; \(RL_1\) and \(RL_2\) = reliabilities of traits 1 and 2; and \(r_{1,2}\) = correlation between PTA for traits 1 and 2. Standard errors (±SE) of the approximate genetic correlations were estimated as:

\[
SE = \sqrt{\frac{1 - \hat{r}_{g1,2}^2}{n - 2}},
\]

where \(n\) = number of bulls with records (Sokal and Rohlf, 1995). Genetic trends were estimated by regression of PTA of cows on birth year.

**RESULTS AND DISCUSSION**

**RT Heritability Estimates**

Genetic and total variance of RT was 0.06 ± 0.03 and 0.28 ± 0.07, respectively (estimate ± standard error). The heritability of RT was estimated to be 0.17 ± 0.13, which is similar to the low end of the range of estimates of Seath (1947; 0.15 to 0.31). The latter study involved analysis of records from the 8 warmest days on which RT were recorded on 99 lactating Holsteins and Jerseys over 2 yr in the summer and autumn (year 1; \(h^2 = 0.15\)) or summer (year 2; \(h^2 = 0.31\)). Thus, the large increase in milk yield of dairy cows and changes in housing conditions since the 1940s does not appear to have altered the heritability of RT substantially. The heritability estimate of RT is greater than or similar to that of several traits currently evaluated, such as DPR (0.04), PL (0.08), and SCS (0.12), which account for 43% of the relative emphasis in lifetime net merit (Cole et al., 2010). The determination of how much weight RT should receive in selection objectives will depend on the economic value of RT and its genetic correlation with other traits, rather than the magnitude of its heritability. Moreover, the large standard error of the estimate of heritability means that accuracy of the estimate could be improved by analysis of more records.

**Genetic Trend for RT**

Genetic trend for RT was estimated by regression of PTA for RT on birth year. Because so few data on RT were available for cows before the birth year of 2002 (4 to 11 observations/yr), results do not include observations before January 1, 2002. Predicted transmitting abilities for RT increased between 2002 and 2008 at an average rate of 0.0068 ± 0.0020°C/yr.

**Genetic and Phenotypic Correlations**

Approximate genetic correlations between RT and 305-d milk, fat, and protein yields, PL, and NM$ were significant and positive, whereas those between RT and SCS and DPR were significant and negative (Table 2). Thus, selection of reduced RT could have adverse effects on genetic merit for production traits, NM$, PL.
and SCS while having beneficial effects on DPR. That genetic merit for RT is adversely related to genetic merit for yield traits is not surprising given that cows that produce more milk generate more heat (Holter, 1976). Similarly, the magnitude of the decrease in fertility caused by heat stress is related to the magnitude of hyperthermia (Gwazdauskas et al., 1973) so it is logical that cows with genetic merit for higher RT would have lower genetic merit for DPR. The only surprising result was the negative genetic correlation between RT and SCS as it implies that genetic merit for high RT reduced inflammation in the mammary gland and, by inference, mastitis. This correlation, which was small, is unexpected because the incidence of mastitis can be greater during summer (Hogan et al., 1989; Elvinger et al., 1991; Makovec and Ruegg, 2003), and heat stress has been reported to decrease the migratory response of PMNL in the mammary gland (Elvinger et al., 1992).

Phenotypic correlations among RT and other traits are shown in Table 3. We observed significant and positive correlations of RT with fat and protein yields, and significant and negative correlations of RT with SCS and PL. Correlations of RT with milk yield and DPR were not statistically different from zero.

To the extent that RT is a proxy for an animal’s ability to resist heat stress, it should be possible to select cows that are more thermotolerant. This conclusion is consistent with that of Oseni et al. (2004), who reported genetic variation in thermal resistance with respect to days open. Genetic selection for RT should be performed using methods that prevent selection for undesirable genes affecting production traits, NM$, and SCS, such as a restricted selection index (Cunningham et al., 1970). A more sophisticated approach would combine information about genetic markers associated with RT but not other traits with PTA from all traits into a selection criterion (Lande and Thompson, 1990). Given the array of molecules involved in regulation of body temperature and cellular responses to hyperthermia, it is likely that many genes contain allelic variants that affect thermotolerance. One such SNP that determines heat stress resistance has been identified in $ATP1A1$ in Holsteins (Liu et al., 2011) and likely, other such genes will be identified.

The heritability estimate for RT had a large error and approached zero at the lowest part of the range. Work is currently underway to collect additional phenotypes from a broader range of herds, which will permit the more precise estimation of genetic correlations. The economic value of RT is also being studied to determine how it might best be incorporated into selection objectives. Nonetheless, data indicate that genetic variation for RT exists in Holsteins and it may be feasible to select for decreased RT during summer months to improve resistance to heat stress. Doing so should be carried out using methods that do not result in inadvertent selection against traits contributing to the economic performance of dairy cows.

**ACKNOWLEDGMENTS**

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**Table 2.** Approximate genetic correlations (±SE) of rectal temperature with other traits

<table>
<thead>
<tr>
<th>Trait</th>
<th>Genetic correlation</th>
<th>SE</th>
</tr>
</thead>
<tbody>
<tr>
<td>305-d Milk yield</td>
<td>0.090*</td>
<td>0.006</td>
</tr>
<tr>
<td>305-d Fat yield</td>
<td>0.096*</td>
<td>0.006</td>
</tr>
<tr>
<td>305-d Protein yield</td>
<td>0.102*</td>
<td>0.006</td>
</tr>
<tr>
<td>305-d SCS</td>
<td>−0.010*</td>
<td>0.007</td>
</tr>
<tr>
<td>Productive life</td>
<td>0.061*</td>
<td>0.006</td>
</tr>
<tr>
<td>Daughter pregnancy rate</td>
<td>−0.013*</td>
<td>0.007</td>
</tr>
<tr>
<td>Net merit</td>
<td>0.122*</td>
<td>0.006</td>
</tr>
</tbody>
</table>

*P < 0.05.

**Table 3.** Phenotypic correlations with rectal temperature and production traits

<table>
<thead>
<tr>
<th>Trait</th>
<th>Correlation</th>
</tr>
</thead>
<tbody>
<tr>
<td>305-d Milk yield</td>
<td>−0.016</td>
</tr>
<tr>
<td>305-d Fat yield</td>
<td>0.067**</td>
</tr>
<tr>
<td>305-d Protein yield</td>
<td>0.090***</td>
</tr>
<tr>
<td>305-d SCS</td>
<td>−0.077***</td>
</tr>
<tr>
<td>Productive life</td>
<td>−0.089*</td>
</tr>
<tr>
<td>Daughter pregnancy rate</td>
<td>0.032</td>
</tr>
</tbody>
</table>

*P < 0.05; **P < 0.01; ***P < 0.001.
REFERENCES


