

Estimation of genetic parameters of the productive and reproductive traits in Ethiopian Holstein using multi-trait models

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Submitted Mar 13, 2017; Revised May 3, 2017; Accepted Jun 13, 2017

Objective: This study estimated the genetic parameters for productive and reproductive traits. **Methods:** The data included production and reproduction records of animals that have calved between 1979 and 2013. The genetic parameters were estimated using multivariate mixed models (DMU) package, fitting univariate and multivariate mixed models with average information restricted maximum likelihood algorithm.

Results: The estimates of heritability for milk production traits from the first three lactation records were 0.03±0.03 for lactation length (LL), 0.17±0.04 for lactation milk yield (LMY), and 0.15±0.04 for 305 days milk yield (305-d MY). For reproductive traits the heritability estimates were, 0.09±0.03 for days open (DO), 0.11±0.04 for calving interval (CI), and 0.47±0.06 for age at first calving (AFC). The repeatability estimates for production traits were 0.12±0.02, for LL, 0.39±0.02 for LMY, and 0.25±0.02 for 305-d MY. For reproductive traits the estimates of repeatability were 0.19±0.02 for DO, and to 0.23±0.02 for CI. The phenotypic correlations between production and reproduction traits ranged from 0.08±0.04 for LL and AFC to 0.42±0.02 for LL and DO. The genetic correlation among production traits were generally high (>0.7) and between reproductive traits the estimates ranged from 0.06±0.13 for AFC and DO to 0.99±0.01 between CI and DO. Genetic correlations of productive traits with reproductive traits were ranged from –0.02 to 0.99.

Conclusion: The high heritability estimates observed for AFC indicated that reasonable genetic improvement for this trait might be possible through selection. The h^2 and r estimates for reproductive traits were slightly different from single versus multi-trait analyses of reproductive traits with production traits. As single-trait method is biased due to selection on milk yield, a multitrait evaluation of fertility with milk yield is recommended.

Keywords: Genetic Parameter; Multi-trait Model; Holstein Friesian; Productive and Reproductive Traits

INTRODUCTION

During the last decades breeding program of dairy herds in Ethiopia had focused on meeting the ever-increasing demand for milk and milk products of the ever increasing population growth in Ethiopia and also contribute to poverty alleviation. Genetic improvement of the indigenous cattle had prioritised on crossbreeding backed by within country selection program on Holstein Friesian herds at Holetta bull dam farm and other follower herds to ensure young bull replacement for artificial insemination services in the country. But, a breeding plan and implementation of selection program lacks detailed genetic parameters analysis on important milk production and reproduction traits like lactation milk yield (LMY), lactation length (LL), age at first calving (AFC), days open (DO), and calving intervals (CI) etc. Therefore, an accurate estimation of genetic parameters in tropical herds is important for planning and implementing efficient breeding programs [1].

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Genetic parameter estimates of locally available exotic and crossbred dairy cattle in Ethiopia have so far been carried out using univariate models. Evaluations based on single trait models does not account for the covariances among traits. Therefore, using estimates from univariate models that do not account for the relationship among traits may result in inaccurate estimates of breeding values that will leads to selection bias. To date multivariate analyses are generally used to select the best replacment stock so that the effect of selection for milk production would be known to affect reproduction traits in the anticipated direction. In additon, using multivariate analyses has a great importance in providing reliable and unbiased esimates of geneic parameters [2]. Variance covariance estimates from a multivariate data analysis in the dairy herd helps to assess the magnitude of genetic correlations among breeding goal traits that enables the setting up of total merit index for the accurate evaluation of the genetic merit of animals in the herd. In general, information on the estimates of genetic parameters of productive and reproductive traits of Holstein Friesian dairy cattle particularly those with multivariate animal models are scarce in tropical environments. Therefore, the objective of this study was to estimate genetic parameters for productive and reproductive traits fitting multi- trait model that would contribute to the design of the national dairy genetic improvement programs in tropical dairy production systems.

MATERIALS AND METHODS

Animals and phenotypes

Data from 1,567 animals of Holstein Friesian born to eight sire origins and 136 sires, 1,431 cows that calved between 1979 and 2013 were used. The data was edited in such a way that incomplete lactation record on lactation milk yield less than 1,000 kg and lactation lengths less than 220 days and greater than 450 days were excluded to avoid bias Ayied et al [3]. In addition, all lactation length records between 21 and 320 days were standardized to 305-days milk yield using the projection procedures of Rege [4]. Lactations greater than 321 days were adjusted by a regression model that considered coefficient of determination and variance inflation factor for each independent variable in the model as described by Khan et al [5]. All animals with AFC between 18 to 60 months were considered as normal expected age range and those animals with records greater than 60 months age for AFC were considered in the terminal category, similarly calving interval between 300 and 900 days and days open between 21 and 500 days were included in the final analysis [6]. Finally a total of 3,552, 3,733, 2,938, 1,125,2,764, and 2,773 records on LMY, 305 days milk yield (305-d MY), LL, AFC, CI, and DO, respectively were used in the final analysis of the mixed models fitted to generate the estimates (Table 1).

Seasons of calving was classified into three groups based on weather and climatic conditions of the area. These included June to September as long rainy season, March to May as short rainy

Table 1. Statistical description on the data set

N, number of observations; SD, standard deviation; 305-d MY, 305 days milk yield; LMY, lactation milk yield; LL, lactation length; AFC, age at first calving; CI, calving interval; DO, days open.

season and October to February as dry season [7]. Further, lactations were classified into three parities as $1, 2$, and $3^{\text{*}}$. Party three and above were all pooled together as parity three due to very few number of observations in later lactations as well as due to the higher correlation between 3rd and later lactations. Sires used in the farm were assigned into eight groups based on their source or country of origin. i.e. those imported from Cuba, Finland, Kenya, Israel, Italy, United States of America, and the remaining recruited in Holetta and unknown group from Ethiopia.

Statistical analysis

Estimation of variance and (co)variance components were made using the multivariate mixed model program package (DMU) using average information restricted maximum likelihood algorithm [8]. To estimate variance and (co)variance components and correlations, single and multi-trait animal models were used for the different traits depending on the nature of the data. A full multi-trait animal model for six traits was initially tried, but it had convergence problems. Other multiple-trait analyses were tried by reducing the number of traits step wisely one at a time from 6 but only the 2 trait and 3 trait multi-trait models resulted in logical estimates. Preliminary analysis was performed to identify the fixed effects which should be considered in the genetic analysis. The presence of any significant differences was checked by using Tukey Kramer multiple comparison tests. Fixed effects which were significant (p<0.05) were fitted in to the model to estimate the genetic parameters. Two different models, model 1 and 2 as shown below were used for estimation of variance components. In general, the fixed effects included in these models were season of calving, year of calving/birth, origin of sire and parity whilst permanent environment, animal genetic and residuals as random effects. The description of the models in matrix notation was:

Model 1 is used for univariate analysis for AFC

 $Y = Xb + Za + e$

Where, Y is the vector of observation; b includes vector of fixed effects; a is the vector of solutions for the coefficients of direct animal (additive) genetic random effects; e is the vector of

the products of additive genetic standard deviations of the two traits whilst phenotypic correlations were

residual effects; and X and Z are incidence matrices of the fixed tesidian encels, and x and z are mederice matrices of the txed
effects and additive genetic random effects, respectively.

Model 2 was used for univariate and multivariate (two trait and three trait) analysis of productive and reproductive traits
(LMY LL 305-d MY CL and DO) fitting permanent environ-(LMY, LL, 305-d MY, CI, and DO) fitting permanent environmental effect due to repeated records per cow.

<u>were the</u> uctive traits
:nt environ-² σ $\overline{}$

 $Y = Xb + Za + Wpe + e$

Where, Y is the vector of observations; b includes vector of where, it is the vector of observations; b includes vector of σ_{di}^2
fixed effects; a is the vector of solutions for the coefficients of direct animal (additive) genetic random effects; pe is the vector phenotypic phenotypic of solution for permanent environmental effects; e is the vector of residual effects; and X, Z, W, are the correspondent incidence
trait *i* and *i* matrices of the fixed effects and additive genetic and permanent environmental random effects, respectively.

In the two trait analysis we considered 305-d MY as a major **RESULTS** trait to evaluate the milk yield of dairy cows as in most modern dairy farms, a lactation length of 305 days commonly accepted as a standard. This standard allows for ideal calving every 12 **Heritability** estimates *Heritability* estimates months with a 60-day dry period. Therefore, we fitted lactation length and reproductive traits with 305-d MY and presented the heritability estimates of all traits in the two trait model. We ex-
The stability for LMY indicated presence of modest gene pected a different h^2 estimate for 305-d MY. However, we found that all the estimates were similar and we just picked the estimates from one of the runs. Likewise, in the three trait model analysis and annual under the study mudated the pesti-
relation among the traits that amplified the herit we considered 305-d MY as a major trait and AFC as second trait based on its h^2 estimate (genetic variability) from the univariate $\frac{f}{f}$ are current estimate from the univariate model is c analysis and economic value of the traits as indicated by Emana [9]. So we have made a combined analysis of 305-d MY and AFC On the other hand, it is higher than those reported with LL, CI, and DO and we got h^2 estimates for all traits just like two trait analysis. In both two and three trait we used LMY instead of 305-d MY merely to test its association with production and fertility traits.

Heritability was estimated as the ratio of the additive genetic
traits models did not show much charvariance to total phenotypic variance; and repeatability, as the ratio of the sum of the additive genetic variance and permanent adjustment made on lactation days that trig environmental variance to phenotypic variance [10]. We was the current estimates from all the current estimates

$$
h^2 = \frac{\sigma_a^2}{\sigma_a^2 + \sigma_{pe}^2 + \sigma_e^2}
$$

$$
r = \frac{\sigma_a^2 + \sigma_{pe}^2}{\sigma_a^2 + \sigma_{pe}^2 + \sigma_e^2}
$$

Where, h⁻ is heritability; r is repeatability; σ_a^2 is direct additive
genetic variance; σ_{pe}^2 is permanent environmental variance revariation within Where, h² is heritability; r is repeatability; σ_a^2 is direct additive national continuous σ_a^2 is now an anti-proposed and variance σ_a genetic variance; $\sigma_{\bar{p}e}$ is permanent environmental variance re-
lated to repeated records and σ_e^2 is residual variance.

The genetic correlations were estimated as the ratio of additive additive covariance of the two traits (trait i and j) to the products of addi-
tion and involuntary culling of animal correlations were estimated similarly as ratio of the sum of the and genetic variability within the populations tive genetic standard deviations of the two traits whilst phenotypic

es of the fixed genetic and environmental covariances (phenotypic covariances) to the products of phenotypic standard deviations of the two traits. traits.

$$
r_g = \frac{\sigma_{aij}}{\sqrt{\sigma_{ai}^2 \sigma_{aj}^2}} \qquad \qquad r_p = \frac{\sigma_{pi}}{\sqrt{\sigma_{pi}^2 \sigma_{pj}^2}}
$$

 σ_{ai}^2 is additive genetic variance for trait *i*; σ_{aj}^2 is additive genetic ects; pe is the vector
phenotypic variance for trait j; σ_{aj} is additive genetic covariance
Forta e is the vector e is the vector
lent incidence between trait *i* and *j*; and σ_{pi} is phenotypic covariance between Where, r_e is genetic correlations; r_p is phenotypic correlation; α is the vector variance for trait *j*; σ_{pi}^2 is phenotypic variance for trait *i*, σ_{pj}^2 is $\frac{1}{s}$ is genere correlations, r_p is prency precorrelations, ² isphenotypic variance for trait *j*; trait *i* and *j*.

RESULTS AND DISCUSSION RESULTS AND DISCUSSION RESULTS AND DISCUSSION

Heritability estimates

The heritability estimates for LMY in the present study from $\frac{1}{\sqrt{1-\frac{1}{\sqrt{1-\frac{1}{\sqrt{1-\frac{1}{\sqrt{1-\frac{1}{\sqrt{1-\frac{1}{\sqrt{1-\frac{1}{\sqrt{1-\frac{1}{\sqrt{1-\frac{1}{\sqrt{1-\frac{1}{\sqrt{1-\frac{1}{\sqrt{1-\frac{1}{\sqrt{1-\frac{1}{\sqrt{1-\frac{1}{\sqrt{1-\frac{1}{\sqrt{1-\frac{1}{\sqrt{1-\frac{1}{\sqrt{1-\frac{1}{\sqrt{1-\frac{1}{\sqrt{1-\frac{1}{\sqrt{1-\frac{1}{\sqrt{1-\frac{1}{\sqrt{1-\frac{1}{\sqrt{1-\frac{1}{\sqrt{1-\frac{1$ resented the two data model was 0.255.0.02 (Table 2). These values of next
del We ex-
tability for LMY indicated presence of modest genetic variability ected a different in estimate for 505-d MT. However, we found
hat all the estimates were similar and we just picked the estimates estimates in two trait analyses for LMY and 305-d MY record of an animal under the study indicated the presence of high coras second trait
the univariate The current estimate from the univariate model is comparable to mode of the runs. EKEWISE, in the three trait moder analysis
considered 305.4 MV as a major trait and AEC as second trait relation among the traits that amplified the heritability estimates. ever, we found among the small population in Holetta. The increment in the ted by Emana those estimates reported for Holstein Friesian dairy cattle [11-13]. the two trait model was 0.25 ± 0.02 (Table 2). These values of heri-
presented the two trait model was 0.25 ± 0.02 (Table 2). These values of herisingle and three trait models were similar (0.17±0.04) whilst in On the other hand, it is higher than those reported for first lactation Holstein Friesians in Egypt [14].

Estimation of genetic parameters the small population in the stud 6 6 The heritability estimates for adjusted 305-d MY from unitimation of genetic parameters the ratio of the small population in the study. However, unlike the estimates to the study of the study of the study of the study. However, unlike the estimates riance re-
variation within population, management and environmental $\frac{1}{2}$ adjustment made on lactation days that triggers weak correlation parable with 0.11 for Egyptian Holstein Friesian dairy cattle [14]. expected and are mainly results of size of dataset, genetic mates are expected and are mainly results of size of dataset, genetic 2 environments [4,16,17]. Such differences in the heritability estif addi-
tion and involuntary culling of animals from the existing herd 6 305-d MY indicate presence of modest genetic variability among $\sigma^2 + \sigma^2$ Similarly, closely heritability estimate of 0.19 reported for the same conditions and the methods used for parameter estimation. In ic might have contributed to the decreased hereditary differences environmental variance related to repeated records and ² is residual variance. variate, 2 trait and 3 trait analysis were 0.15±0.04, 0.16±0.04, and 0.15±0.04, respectively (Table 2). These values of heritability for for LMY the heritability estimates in univraite, two and three traits models did not show much change, this could be due to to other traits. The current estimates from all models were combreed in a temperate environment [15]. However, higher estimates were reported by several other authors from different production addition, we believed that a long term biased phenotypic selecand genetic variability within the population. Although estimate

Table 2. Estimates of permanent environmental (σ²_β), additive genetic (σ²₈), and residual variances (σ²₈), heritability (h²) and repeatability (r) of production and reproduction traits using univariate, two and tree trait models

Models	Traits	$\sigma^2_{~\text{pe}}$	$\sigma^2_{\ a}$	$\sigma^2_{\ e}$	$h^2 \pm SE$	$r\pm$ SE
Univariate model	305-d MY	287,808.6	160,216.1	646,915.1	0.15 ± 0.04	0.42 ± 0.02
	LMY	294,985	229,963	830,492	0.17 ± 0.04	0.39 ± 0.02
	LL	181.8	60.3	1,834.3	0.03 ± 0.03	0.12 ± 0.02
	AFC		22.7	25.9	0.47 ± 0.06	
	CI.	1,526.5	1,483.02	10,965	0.11 ± 0.04	0.22 ± 0.02
	D ₀	1,020.0	982.6	8,571.5	0.09 ± 0.03	0.19 ± 0.02
Two traits model	305-d MY	291,396.8	195,089.6	738,703.6	0.16 ± 0.04	0.41 ± 0.02
	LMY	174,216.2	273,808.6	644,915.1	0.25 ± 0.02	0.45 ± 0.02
	LL	116.5	136.6	1,834.4	0.06 ± 0.03	0.12 ± 0.02
	AFC		18.2	29.7	0.38 ± 0.27	
	^C	1,526.0	1,678.2	10,927.8	0.11 ± 0.04	0.23 ± 0.02
	D ₀	1,006.5	1,107.0	8,558.8	0.09 ± 0.03	0.20 ± 0.02
Three traits model	305-d MY	287,808.6	160,216.1	646,915.1	0.15 ± 0.04	
	LMY	229,781	294,639	830,914	0.17 ± 0.04	
	LL.	116.5	166.6	1,804.4	0.08 ± 0.03	
	AFC	1,767.7	1,436.8	1,0954.7	0.38 ± 0.25	
	\overline{C}		32.5	52.3	0.13 ± 0.04	
	D ₀	1,080.2	1,017.2	8,579.2	0.10 ± 0.04	

SE, standard error; 305-d MY, 305 days milk yield; LMY, lactation milk yield; LL, lactation length; AFC, age at first calving; CI, calving interval; DO, days open.

of heritability for LMY and adjusted 305-d MY indicates that still there is an opportunity for improving these traits through selection, future improvement of the herd could be meaningful if the current phenotypic selection is changed to selection based on genetic merit combined with improved herd management as well as introducing of new animals from other source to increase genetic variability within the herd.

The heritability estimates for lactation length ranged from 0.03±0.03 to 0.08±0.03 (Table 2) and these estimates correspond to those reported by Ojango and Pollott [17]. It is slightly higher than heritability estimates of 0.003±0.08 in Sudan using paternal half-sib correlation [18]. These low heritability estimates obtained in this study could be attributable to the high residual variance. In the study the heritability estimates of LL shows slight improvement from univariate model to 3 traits model. This enhanced accuracy is due to the ability of the model to account for relationship between the traits and better connections in the data due to residual covariance between traits [19]. Moreover, multivariate models gave better reliabilities than univariate models in the prediction of genetic merit which is mainly a result of the model's ability to use extra information from correlated traits [20].

Heritability estimates for AFC were 0.47, 0.38, and 0.38 from univariate, 2 trait and 3 trait models, respectively (Table 2). These high heritability estimates for AFC from this study may be explained by weak selection made on this trait in contrast to the intense selection practiced for milk production and presence of genetic variability in the small herd as the result of the different sources of sires being used in the study period. The current estimates were comparable with estimates of Rege [4] and Ojango and Pollott [17] who reported 0.47 and 0.38, respectively for the

same breed in Kenya. Differences between the estimates of heritability obtained in the study and estimates from other countries are most likely caused by management and climate differences affecting genetic and environmental variances and methods of estimation [21]. Our study suggests that improvement of this trait through selection in line with several other authors who have come to the same conclusion [4,22].

Heritability estimates for calving interval were 0.11±0.04, 0.11±0.04, and 0.13±0.04 from univariate, 2 trait and 3 trait models, respectively (Table 2). These estimates for CI is still within the literature estimates for the traits, indicating the existence of certain level of genetic diversity. The present result is lower than estimates of 0.17 for Iraqi Friesian cows [13]. However, it is higher than the estimates reported for the same breed in other tropical countries [17,23,24]. This relatively low heritability estimates for CI could be explained by large environmental variance. Therefore, improvements in nutrition and reproductive management should lead to a considerable decrease in length of CI than making purely genetic selection alone. However, in the study, multi-trait analysis with milk production trait and AFC indicates slight improvement in the estimates of heritability of calving interval. Incorporating information on correlated traits may enhance the accuracy of selection and reduce selection bias made by univariate analysis [22]. Therefore, a multiple trait model selection including milk production with fertility traits could improve the selection accuracy for selecting the right animals with good fertility traits [15,25].

Heritability estimates for days open in the same study were 0.09±0.03, 0.09±0.03, and 0.10±0.04 from univariate, 2 trait and 3 trait models, respectively (Table 2). The heritability estimates from all models were similar and coincide with most literatures.

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The heritability estimates for DO is also indicative of the presence of genetic basis that deserve selection or culling of an animal. The current estimates were close to 0.082 reported by Zambrano and Echeverri [24], but higher than 0.042 that was reported in temperate zone [15]. On the contrary, higher estimates of days open have been reported in Egypt and Iraq for the same breed [26,13].

Repeatability estimates

In the current study the repeatability estimate for LMY and adjusted 305-d MY is moderate (Table 2). For LMY the repeatability estimates were 0.39 and 0.45 form univariate and 2 trait models, respectively, which are comparable with that reported by Ojango and Pollott [17], but higher than the reported value of 0.26±0.08 for the same breed in Pakistan [12]. As for adjusted 305-d MY the repeatability estimate were 0.42 and 0.41, these values were in close agreement with 0.43 reported by Dematawewa and Berger [15]. Nevertheless, Rege [4] and Kadarmideen et al [16] have reported higher repeatability estimates of 0.55 and 0.58, respectively. These moderate repeatability estimates in the present study indicated that selection of dairy cattle for milk yield based on early performance records is reliable.

Regarding lactation length the repeatability estimate was very low which correspond with Ojango and Pollott [17]. However, slightly higher estimate of 0.19 was reported for the same breed [12]. The low repeatability estimate implied that variation in lactation length was more a result of variation in management and feeding in the given environment of a particular lactation rather than of factors associated genetic makeup of the cow [17,12].

The repeatability estimates for calving interval and days open from all models were very low (Table 2). Estimates for CI and DO were comparable with Rege [4] who reported 0.25 for the same breed in Kenya, but higher than estimates reported by Zambrano and Echeverri [24]. The low repeatability estimates obtained in this study suggest that CI and DO are strongly influenced by temporary environmental factors, this is mainly because of the complex nature of reproductive traits, difficulties in detection of oestrus, and various other managerial and nutritional factors.

Genetic correlations

Genetic correlations between milk production traits were high. The highest genetic correlation was 0.99±0.01 between LMY and 305-d MY and the lowest was 0.73±0.19 between 305-d MY and LL (Table 3). This is comparable with Ahmad et al [27]. The highest genetic correlation between milk productions traits were result from pleiotropy [10]. This means that genetic improvement of one milk production trait could result in a correlated response in the correlated trait. Generally the very high correlation between these two traits is indicative of the fact that using 305-d MY can be sufficient to meet selection for lactation milk yield.

The genetic correlation between 305-d MY and AFC was –0.24±0.11 (Table 3). This is in close agreement with –0.22±0.08 reported in Ethiopia [28]. On the contrary, a relatively higher and positive (0.54) genetic correlation was reported for Holstein Friesian dairy cattle in Kenya [17]. The negative genetic correlation between 305-d MY and AFC indicates that genes which affect the 305-d MY trait positively will also be responsible for an early onset of puberty among the current population found at Holetta bull dam herd.

Similarly genetic correlation between 305-d MY and CI found to be –0.10±0.15. This result is accords with–0.03 correlation estimate of Friesian cattle in Kenya [4]. However, higher and positive genetic correlations of 0.54±0.01 was reported for the same breed in Ethiopia [28]. Regarding genetic correlation between 305-d MY and DO, it was close to zero (–0.02±0.16). Positive and lower estimates were reported in Turkey [29] and Egypt [14]. On the other hand, higher and positive estimates of genetic correlation was observed for Ethiopian Holstein [28]. The negative genetic correlation of 305-d MY and fertility traits indicates that improvement of fertility is possible to certain degree in this herd as milk yield increase. However, since the standard error of the estimates were high, the results needs to be interpreted with caution. The higher standard errors obtained from pair wise combinations of CI and DO with 305-d MY could be due to the smaller proportion of CI and DO records in the dataset.

In the current study, positive genetic correlations between various reproductive traits were observed. This positive genetic correlation between reproductive traits indicates that improvement of one trait has positive impact on the other trait. The genetic correlation between CI and DO was 0.99±0.01 (Table 3). This is in close agreement to a nearly perfect genetic correlation of 0.99 [30] and 0.98 [24]. The positive and strong genetic correlation in the present study indicate the evidence of common genetic and physiological mechanisms controlling those traits. In the tropics in general and in the Holetta bull dam herd in particular

Table 3. Genetic (above the diagonal) and phenotypic (below the diagonal) correlations between milk production and fertility traits using two trait analysis

SE, standard error; 305-d MY, 305 days milk yield; LMY, lactation milk yield; LL, lactation length; AFC, age at first calving; CI, calving interval; DO, days open.

the recording system is still at its early stage of development. As a result, recording of calving interval is relatively easier than days open. Therefore, evaluation based on calving interval would be easy to handle by the available human resource and recording infrastructure than evaluation based on days open.

Phenotypic correlation

In this study phenotypic correlations between milk productions traits were high (Table 3). The highest phenotypic correlation was 0.84±0.01 between 305-d MY and LMY and the lowest was 0.05±0.02 between 305-d MY and LL. This explains that these traits are influenced by more or less same genetic and environmental factors. The highest phenotypic correlation among LMY and 305-d MY was in agreement with the estimate reported for crossbreeds dairy cattle in Egypt [26,27].

Phenotypic correlations between 305-d MY with fertility traits (AFC, CI, and DO) were in general negative (Table 3). The phenotypic correlation between 305-d MY and AFC in present study was higher than Ahmad et al [27] for crossbreed cattle (–0.02) and –0.02 for first parity Holstein Friesian [17]. Conversely, positive phenotypic correlation was reportedfor Jersey cattle in Ethiopia [28]. In the current study the phenotypic correlation between CI and DO is 0.98±0.01 (Table 3). This strong and positive phenotypic correlations is in agreement with results of Kadarmideen et al [25], Ghiasi et al [30], and Zambrano and Echeverri [24] who reported 0.95, 0.95, and 1.0 for Holstein Friesian cattle in different production environment, respectively.

CONCLUSION

Estimates of genetic parameters are essential strategic step in launching a breeding programs which could be used as a tool for implementation of selective breed improvement in a population. Therefore, the correct estimation of genetic parameters that guarantee for accurate prediction of genetic merit of an individual is essential. The heritability estimates for all production and reproduction traits indicated that the major part of variation for those traits were due to environmental factors. Although the heritability estimate for LMY and 305-d MY were relatively low, there is still some room for genetic improvement in the herd. The repeatability estimate for lactation length and all reproductive traits were low indicating that these traits are mainly influenced by managerial and temporary environmental effects. The genetic correlations observed for milk production and reproductive traits ranged from moderate to high. Multi-trait analysis also showed slight improvement in heritability estimates of CI and LL than univariate model analysis. This indicates that selection with multitrait model could be better than univariate model due to the correction of selection bias by multivariate model as a result of using the information from the correlated traits. Therefore, in the future verification of some lower estimates of certain traits including large set of data and fitting multivariate model for both

CONFLICT OF INTEREST

We certify that there is no conflict of interest with any financial organization regarding the material discussed in the manuscript.

ACKNOWLEDGMENTS

The authors are grateful for the technical and financial support by Jimma University, Ethiopia and the Natural Resource Institute of Finland (Luke). The National Artificial Insemination Centre (NAIC), Ethiopia and particularly the management Holetta dairy farm are duly acknowledged for allowing us to use their farm record.

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