Potential influence of κ-casein and β-lactoglobulin genes in genetic association studies of milk quality traits

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Objective: From a review of published information on genetic association studies, a meta-analysis was conducted to determine the influence of the genes κ-casein (CSN3) and β-lactoglobulin (LGB) on milk yield traits in Holstein, Jersey, Brown Swiss, and Fleckvieh.

Methods: The GLIMMIX procedure was used to analyze milk production and percentage of protein and fat in milk. Models included the main effects and all their possible two-way interactions; not estimable effects and non-significant (p>0.05) two-way interactions were dropped from the models. The three traits analyzed used Poisson distribution and a log link function and were determined with the Interactive Data Analysis of SAS software. Least square means and multiple mean comparisons were obtained and performed for significant main effects and their interactions (p<0.0255).

Results: Interaction of breed by gene showed that Holstein and Fleckvieh were the breeds on which CSN3 (6.01%±0.19% and 5.98%±0.22%), and LGB (6.02%±0.19% and 5.70%±0.22%) have the greatest influence. Interaction of breed by genotype nested in the analyzed gene indicated that Holstein and Jersey showed greater influence of the CSN3 AA genotype, 6.04%±0.22% and 5.59%±0.31% than the other genotypes, while LGB AA genotype had the largest influence on the traits analyzed, 6.05%±0.20% and 5.60%±0.19%, respectively. Furthermore, interaction of type of statistical model by genotype nested in the analyzed gene indicated that CSN3 and LGB genes had similar behavior, maintaining a difference of more than 7% across analyzed genotypes. These results could indicate that both Holstein and Jersey have had lower substitution allele effect in selection programs that include CSN3 and LGB genes than Brown Swiss and Fleckvieh.

Conclusion: Breed determined which genotypes had the greatest association with analyzed traits. The mixed model based in Bayesian or Ridge Regression was the best alternative to analyze CSN3 and LGB gene effects on milk yield and protein and fat percentages.

Keywords: Dairy Cattle; Genetic Improvement; Polymorphism

INTRODUCTION

Genetic association studies have been increasingly used in cattle breeding programs. However, the results have been inconsistent for milk protein genes. Positive, negative, or absence of association with similar genotypes have been reported [1]. Examples of these are the conclusions of Duifhuis-Rivera et al [2] and Dogru [3] who reported that different κ-casein (CSN3) and β-lactoglobulin (LGB) genotypes were not associated with milk yield in a Mexican herd of Holstein and in a Turkish herd of Brown Swiss cattle, respectively. On the contrary, Gustavsson et al [4] reported that the composite genotype BB/A1A2/AB of CSN1S1/CSN2/CSN3 has positive effects on cheese yield and percentage of protein and fat in milk.

The inconsistencies have been attributed to various issues affecting the production and composition of milk. Bernabucci et al [5] concluded that the sampling station affects total protein
concentration and acidity in milk, ranging from 3.2% to 6.0% and from 2.0% to 5.6%, respectively. Moreover, Streit et al. [6] documented the importance of determining the alleles in the diacylglycerol O-acyltransferase 1 genotype in German Holstein sires to avoid differences that can be as high as 20% in the association of these alleles with milk production values as a result of the allele substitution effect.

However, there is no accurate information related to the potential influence of CN3 and LGB genes that could be used in genetic association studies to analyze samples obtained under different racial conditions and sampling methodologies. Therefore, in this study we conducted a meta-analysis of genetic association studies to determine the behavior of genes CN3 (Genbank Accession No. AC_000163.1) and LGB (Genbank Accession No. AC_000168.1) on milk yield and percentage of protein and fat. The three traits analyzed were milk production and percentage of milk protein and fat. The random and fixed effects included in the final model were milk production and percentage of milk protein and fat under different conditions and methods of analysis.

**MATERIALS AND METHODS**

**Paper selection criteria**

Scientific journals were searched for published papers on genetic association studies. Initially, one hundred and forty-seven papers were chosen. The used criteria aimed to eliminate their heterogeneity, to look for representability of the results, and to ensure their replicability. In a first step, the papers published from 2003 to 2016 were selected. Subsequently, we considered only studies dealing with the most studied milk protein genes, CN3 and LGB. The remaining papers described studies on milk production and protein and fat percentage in milk. The final sample included 26 papers dealing with Holstein, Jersey, Brown Swiss, and Fleckvieh dairy breeds.

**Meta-analysis**

Traits analyzed were milk production and percentage of milk protein and fat. The random and fixed effects included in the final model were determined by first establishing a complete model by trait. This model included the main effects and all their possible two-way interactions; some of these were not estimable and were dropped from the model. Additionally, non-significant (p>0.05) two-way interactions were deleted as well. The best fit model and link function were determined with the modules Distribution (Y) and Fit (Y, X), respectively; both are modules from the Interactive Data Analysis of SAS software [7]. The three traits analyzed had the best fit with a final model that used Poisson distribution and a log link function. The final models for milk production, protein and fat percentages were as follows:

\[
MP = A_i + B_i + T_x + AG + GN + SS + BAG + BGN + MGN + E
\]

Where MP = milk production; \(A_i\) = random effect of \(i\)-th paper included in the study; \(B_i\) = \(i\)-th effect of the breed; \(M_j\) = \(j\)-th effect of the used model in the original study; \(T_x\) = \(x\)-th effect of the used test to recover the original data; \(AG\) = \(w\)-th effect of the analyzed gene; \(GN\) = \(v\)-th effect of the genotype nested in the analyzed gene; \(SS\) = \(u\)-th effect of the sample size used; \(BAG\) = \(t\)-th effect of the interaction of breed by analyzed gene; \(BGN\) = \(s\)-th effect of the interaction of breed by genotype nested in the analyzed gene; and \(E\) is the residual random effect.

\[
P = A_i + B_i + M_j + T_x + AG + GN + SS + E
\]

Where \(P\) = protein percentage; \(A_i\) = random effect of \(i\)-th paper included in the study; \(B_i\) = \(i\)-th effect of the breed; \(M_j\) = \(j\)-th effect of the used model in the original study; \(T_x\) = \(x\)-th effect of the used test to recover the original data; \(AG\) = \(w\)-th effect of the analyzed gene; \(GN\) = \(v\)-th effect of the genotype nested in the analyzed gene; \(SS\) = \(u\)-th effect of the sample size used; and \(E\) is the residual random effect.

\[
F = A_i + B_i + M_j + T_x + AG + GN + SS + E
\]

Where \(F\) = fat percentage; \(A_i\) = random effect of \(i\)-th paper included in the study; \(B_i\) = \(i\)-th effect of the breed; \(M_j\) = \(j\)-th effect of the used model in the original study; \(T_x\) = \(x\)-th effect of the used test to recover the original data; \(AG\) = \(w\)-th effect of the analyzed gene; \(GN\) = \(v\)-th effect of the genotype nested in the analyzed gene; \(SS\) = \(u\)-th effect of the sample size used; and \(E\) is the residual random effect.

Once all the information from the papers of the sample had been gathered, an adjustment of the information expressed as deviations from their mean was conducted. The population mean reported in each article was used as a reference mean for the analyzed traits. The adjusted values were included in a final database and the statistical analysis was conducted. The GLIMMIX procedure [7] was used to analyze the information. For the main effects and their interactions that were significant (p<0.0255), least square means were obtained and multiple mean comparisons were performed.

The main effects considered in the analysis were the following: article (A), each article included in the study; breed (B), Holstein, Jersey, Brown Swiss, and Fleckvieh; model (M), each of the statistical analysis approaches used in the papers (least square means and mixed models based on Bayesian or Ridge Regression: MMBRR); test (T), the test to recover the data used in the papers (305-day and one-day test); gene (AG), CN3 and LGB; genotype (GN), AA, AB, AE, BB, BE, and BE for CN3, and for LGB were AA, AB, and BB; sample size (SS), size of the population analyzed in the paper, i) 1 to 500 animals was considered small, and ii) from 501 to 2,000 was regarded as medium. The final model for milk production included some two-way interactions: breed by gene; breed by genotype nested in analyzed gene; and model by genotype nested in analyzed gene. The effect of paper was
regarded as random, the rest of them were considered fixed.

RESULTS AND DISCUSSION

Table 1 presents the level of significance for the effects considered in the models to analyze milk production (MP) and protein (PP), and fat percent (FP) in milk. The type of test used was highly significant (p<0.0001) for milk production, probably because the data used in the meta-analysis included one-day tests as well as 305-day tests. On this regard, Gustavsson et al [4] and Poulsen et al [8] concluded that the biases associated with these tests could be controlled by grouping measurements obtained with the same test.

The sample size was significant (p<0.0001) for milk production. On this regard, Vidović et al [9] concluded that bias caused by the sample size is due to the difficulty of differentiating the effects of each gene in the analysis of polygenic traits. On the other hand, the effect of the interaction model by genotype nested in the analyzed gene was significant (p≤0.0255) probably because of the variance explained by the random effects considering the mixed model [9,10].

The interaction of breed by analyzed gene, was highly significant (p<0.0001). The effect of the gene analyzed had different behavior across breeds and between genes within breed (Table 2). The largest difference in CSN3 and LGB was between the Holstein and Jersey breeds, with 7.65% and 7.64% higher in Holstein than in Jersey, respectively. In Fleckvieh and Brown Swiss breeds, both genes are similarly associated with milk production and total solids, showing greater influence of both genes in Fleckvieh, up to 7.65% and 7.64% for CSN3 and LGB, respectively, compared with Brown Swiss.

Deb et al [11], mentioned that breed has a marked effect on milk yield but somewhat less of an effect on milk composition. The heavier breeds tend to produce more milk. Therefore, the breeds with high milk production (Holstein and Fleckvieh) showed the greatest response. On this regard, since the improvement programs in cattle are based on the genetic potential that the individuals could show. These results may be due to selection for traits such as milk production and total solids that are associated with CSN3 and LGB in both breeds [12,13]. Additionally, genetic make-up of dairy animals plays a great role in the variation of milk yield and composition. Raven et al [14] and Ramayo-Caldas et al [15] mentioned the importance of determine the proportion of genetic markers shared between breeds to study and compare them in multi-breed multi-trait association studies in commercial herds. On this regard, Ramayo-Caldas et al [15], reported around 206 genes with the same effect in three French breeds. Meanwhile, Raven et al [14], determined that despite the different linkage disequilibrium patterns, Holstein and Jersey share between 8% to more than 38% genetic markers with similar effects on economic important traits in dairy cattle.

The interaction of breed by genotype nested in the analyzed gene (Table 3) was highly significant (p<0.0001). Holstein and

### Table 1. Level of significance for the effects included in the models to analyze milk production (MP), and protein (PP) and fat (FP) percentage in milk

| Variable | \( B_i \) | \( M_i \) | \( T_i \) | \( \beta_{\text{gen}} \) | \( \kappa_{\text{gene}} \) | \( \alpha_{\text{gen}} \) | \( \kappa_{\text{GEN}} \) | \( \alpha_{\text{GEN}} \) | \( \kappa_{\text{SS}} \) | \( \alpha_{\text{SS}} \) | \( \kappa_{\text{BAG}} \) | \( \alpha_{\text{BAG}} \) | \( \kappa_{\text{BGN}} \) | \( \alpha_{\text{BGN}} \) | \( \kappa_{\text{MGN}} \) | \( \alpha_{\text{MGN}} \) |
|----------|----------|----------|----------|----------------|----------------|----------------|----------------|----------------|----------------|----------------|----------------|----------------|----------------|----------------|----------------|----------------|----------------|
| MP       | 0.1528   | 0.2411   | <0.0001  | 0.1833        | 0.0974         | <0.0001        | 0.0974         | <0.0001        | 0.0974         | <0.0001        | 0.0974         | <0.0001        | 0.0974         | <0.0001        | 0.0974         | <0.0001        | 0.0974         | <0.0001        |
| PP       | 0.7529   | 0.7358   | 0.6484   | 0.9966        | 1.000          | 0.9090         | -              | -              | -              | -              | -              | -              | -              | -              | -              | -              | -              |
| FP       | 0.1512   | 0.4161   | 0.8409   | 0.8604        | 1.000          | 0.8199         | -              | -              | -              | -              | -              | -              | -              | -              | -              | -              | -              |

B, breed; \( M_i \), model used in the original study; \( T_i \), test used to recover original data; \( \beta_{\text{gen}} \), analyzed gene; \( \kappa_{\text{gene}} \), genotype nested in analyzed gene; \( \alpha_{\text{gen}} \), sample size; \( \kappa_{\text{BAG}} \), interaction of breed by analyzed gene; \( \alpha_{\text{BAG}} \), interaction of genotype nested in analyzed gene; \( \kappa_{\text{BGN}} \), interaction of model by genotype nested in analyzed gene; \( \alpha_{\text{BGN}} \), interaction of model by analyzed gene; \( \kappa_{\text{MGN}} \), interaction of model by analyzed gene.

### Table 2. Least square means and multiple comparison for milk production for the interaction breed by gene

| Breed      | Gene   | Least square means, standard error, and multiple comparison for milk production for the subclasses of the interaction breed by genotype in the loci of the CSN3 gene and LGB gene |
|------------|--------|--------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|
|            |        | Genotype of the CSN3 gene | Genotype of the LGB gene |
| Breed      | AA     | AB | BB | AA | AB | BB |
| Brown Swiss | 5.54±0.47\(^a\) | 5.62±0.47\(^a\) | 5.64±0.47\(^a\) | 5.54±0.47\(^a\) | 5.62±0.47\(^a\) | 5.64±0.47\(^a\) |
| Holstein   | 6.04±0.22\(^a\) | 6.01±0.22\(^a\) | 6.00±0.22\(^b\) | 6.05±0.20\(^a\) | 6.01±0.20\(^a\) | 6.01±0.20\(^a\) |
| Jersey     | 5.59±0.31\(^a\) | 5.57±0.31\(^a\) | 5.50±0.31\(^a\) | 5.60±0.19\(^a\) | 5.54±0.19\(^a\) | 5.57±0.19\(^a\) |
| Fleckvieh  | 5.80±0.24\(^a\) | 5.86±0.24\(^a\) | 6.30±0.22\(^a\) | 5.71±0.22\(^a\) | 5.73±0.22\(^a\) | 5.67±0.22\(^a\) |

CSN3, \( \kappa \)-casein; LGB, \( \beta \)-lactoglobulin.

\(^a\) Means with different letter in the same row or column are different (p<0.001).
Jersey had a greater influence of genotype AA in CSN3 gene on milk yield, up 1.61% from other genotypes studied; Miciński et al [12] reported a difference of 0.92% to 11.94% for Polish Jersey.

Estimates for Brown Swiss and Fleckvieh were different from those for Holstein and Jersey; the BB genotype was the most closely associated with milk production, 7.94% higher than the other studied genotypes. Similarly, Chrenek et al [16] found that, for Fleckvieh, the genotype most closely associated with milk yield is BB, 7.42%. In contrast, Matějíček et al [17] concluded that genotype AB was the most closely associated with that trait.

Similar behavior shown in the study by Holstein and Jersey and Brown Swiss and Fleckvieh may also attend mainly the common geographic origin of the analyzed breeds. In this regard, Negrini et al [18] calculated genetic distances through genetic fingerprinting of 51 cattle breeds. They placed the Holstein and Jersey breeds in the Nordic genetic type group, while the Brown Swiss and Fleckvieh were grouped with genetic types from the Alpine region or Central-France.

Holstein and Jersey showed similar behavior for AA genotype of the LGB gene (Table 3). This estimate had the largest influence on milk yield, up to 0.66% and 1.07% above the other genotypes. This result is similar to that found for the CSN3 gene in this study since, according to Bonfatti et al [19], the LGB gene is also associated with milk production and total solids. Fleckvieh and Brown Swiss are influenced in a major way by genotypes BB and AB, showing up to 1.77% and 1.05% difference, relative to other genotypes. On this regard, Gustavsson et al [4] determined the effect of breed and genotype using composite genotypes of κ-casein and β-lactoglobulin genotypes to determine their genetic association with production traits. These authors concluded that the differences between the genetic association values, ranging between 0.5% and 30%, were mainly influenced by genotype. Moreover, Oltenacu and Broom [20], report rising of inbreeding rates of 0.2% per year in Holstein and Jersey which would cause a decreasing in the response to improvement and selection program and the genetic association values due to loss of the genetic variation in the population.

On the other hand, the interaction of type of statistical model used by genotype nested in the analyzed gene was highly significant (p<0.0001), and mean estimates are shown in Table 4.

The differences between the models used for each of the studied genes may be explained with the conclusions of Miciński et al [12] and Monir and Zhu [21], who demonstrated that the inclusion or omission of the effect of each particular gene and polygenic effects, influence the results of production traits and only the models that include random effects are able to differentiate those changes.

The results of comparing models of LGB and CSN3 genes (Table 4) were similar, maintaining a difference of just over 7% across all genotypes. Here, Comin et al [10] and Vidović et al [9] concluded that the MMBRR explained the effect of milk protein genotypes on performance and composition of different breeds of dairy cattle, nearing a 9% difference, relative to studies using only models with fixed effects based on least squares.

Oleriński et al [22] and Pärna et al [23] used models similar to ours when attempted to explain the variability present in genetic association studies. However, the general linear mixed models were not able to differentiate changes between AB and BB genotypes of the LGB gene, while for the CSN3 MMBRR had the best fit. Here, Kučerová et al [24] concluded that the mixed model is the best suited for studies of genetic association with genes of casein; while such models shown less power to determine associations for genes of other milk-whey proteins, including β-lactoglobulin.

**CONCLUSION**

Differences in the magnitude of the influence of CSN3 and LGB genotypes depending on the breed could change according on the shared genetics. The mixed model based in Bayesian or Ridge Regression was the best alternative for analysis in genetic association studies involving the CSN3 and LGB genes. Due to their higher substitution allele effect and the minor inbreeding level, Brown Swiss and Fleckvieh could show more progress in selection programs that include the CSN3 and LGB genes, relative to Holstein or Jersey.

**CONFLICT OF INTEREST**

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