Article

Basal Levels of CD18 Antigen Presenting Cells in Cow Milk Associate with Copy Number Variation of Fc Gamma Receptors

Eyal Seroussi 1,* , Shlomo E. Blum 2,*, Oleg Krifucks 2, Andrey Shirak 1, Shamay Jacoby 1 and Gabriel Leitner 2

1 Institute of Animal Science, ARO, Volcani Center, HaMaccabim Road, P.O. Box 15159, Rishon LeTsiyon 7528809, Israel; shiraka@volcani.agri.gov.il (A.S.); shamijac@gmail.com (S.J.)
2 National Mastitis Reference Center, Kimron Veterinary Institute, P.O. Box 12, Bet Dagan 50250, Israel; shlomobl@moag.gov.il (S.E.B.); olegkr@moag.gov.il (O.K.); leitnerg1@gmail.com (G.L.)
* Correspondence: seroussi@agri.huji.ac.il; Tel.: +972-8-9484425

Received: 20 July 2020; Accepted: 17 August 2020; Published: 18 August 2020

Abstract: Differentiation of cells by flow cytometry provides informative somatic cell counts (SCCs) that allow analyzing leukocyte population patterns in udder infections of different etiologies. Postulating that this approach also enhances the statistical power to detect genetic variants linked to cell levels in milk of healthy mammary glands, we used monoclonal antibodies anti-CD18, anti-CD4, anti–CD14, and anti-PMN to count cells presenting these surface antigens, and performed a genome-wide association study of these counts in 125 Israeli Holsteins genotyped using SNP BeadChips. We identified an informative haplotype of 15 SNPs in the centromeric end of BTA3 that was strongly associated with CD18 cells ($p < 2.3 \times 10^{-9}$). Within this region, examination of the network of genes interacting with ITGB2 (CD18) indicated an Fc-$\gamma$-receptor gene cluster, including FCGR2A (CD32). Sanger-sequence analysis of FCGR2s-linked exon 3 variation to CD18 counts. Meta-analysis of RNA-Seq data revealed a significant negative correlation ($R = -0.51$) between expression of CD32 and CD18 in milk. Assembly of DNA-Seq reads uncovered FCGR copy-number variation and a variant, designated V7, was abundant in dairy cattle, probably reflecting adaptation to selection pressure for low SCC in Holstein milk.

Keywords: SNP BeadChip; cluster of differentiation; classification determinant; GWAS; milk; mastitis; immunogenetics

1. Introduction

Mastitis—inflammation of the mammary gland—results in the most significant economic losses to the dairy industry worldwide [1]. Incidents of mastitis are numerous and can reach up to 75% of the herd in developing countries [2]. Its cost involves drugs, labor, discarded and shorter shelf-life milk, reduced cheese yield, lower price paid for lower milk quality, and increased culling. Cows that are culled or removed from the herd are mostly replaced by first-calf heifers. Lameness, disease, clinical and subclinical mastitis, and mortality force involuntary herd management and disrupt the voluntary management plan of culling old, genetically and/or physiologically low-producing and infertile cows.

Bacteria are the major causative pathogens of mastitis. They are distributed all over the cow’s skin, including the external part of the streak canal, and are part of the flora of the cow’s environment. However, regardless of environmental conditions and management, many cows/glands are not inflamed by mastitis, indicating that genetics may play an important role. Penetration of bacteria into the mammary gland, mostly through the teat canal, is crucial for the development of
mastitis. However, not every bacterium passing through the teat canal will end up causing mastitis. Early elimination of the bacteria by innate immune factors such as phagocytes makes it impossible to calculate the percentage of bacterial elimination. What can be evaluated is the distribution of animals with identified clinical and/or sub-clinical mastitis in the entire population. However, unlike the clinical form, determination of sub-clinical mastitis is problematic due to a lack of visual symptoms. Thus, in most studies related to sub-clinical mastitis, mastitis detection at cow or quarter levels is based on the California mastitis test (CMT) and somatic cell count (SCC), sometimes together with bacteriology (for more details, see [3]). In large genetic analyses of mastitis cases, the major and sometimes only indicator of sub-clinical mastitis is SCC, which is determined monthly or more frequently. However, the relation of these data to mastitis lacks sensitivity and details: (1) in most cases, only one gland is infected and therefore cow SCC is low (<100,000 cells/mL); moreover, in many studies, the threshold of 200,000 cells/mL for a healthy cow is considered; (2) in some clinical cases, the cow is treated and the SCC decreases within days to <200,000 cells/mL; therefore, if the infection occurs between two routine monthly milk recordings, the incidence of mastitis may go unnoticed, unless it is recorded by hand; (3) in many cases, a bacteriology test is not performed and even when it is, identification is at the species level. Therefore, the data for analysis, if recorded, include the severity and duration of the inflammation induced by bacteria, the host’s immunological response, and their interaction. Important parameters of this immunological response are the counts of leukocytes in the host’s milk [4,5]. In this work, we analyzed the association between the genomes of 148 dairy heifers and their udder health with milk cells expressing common cluster of differentiation (CD) antigens, which are cell-surface molecules that provide targets for the immunophenotyping of leukocytes.

2. Materials and Methods

2.1. Animals and Experimental Scheme

The study was conducted in one herd of ~200 Israeli Holstein dairy cows producing an average milk yield of >11,500 L/305 lactation days. All treatment protocols were approved by the Institutional Animal Care Committee of the Agricultural Research Organization, Volcani Center, Bet Dagan, Israel (# 16_b7736_10). The dairy parlor has been previously described [6]. Briefly, it is equipped with a computerized AfiFarm herd-management system and AfiLab milk analyzer, providing on-line data on gross milk composition and conductivity. Cows were milked thrice daily at 04:00, 13:00, 20:00 h. Routine monthly milk yield, milk composition, and SCC were recorded by the Israeli Cattle Breeders Association (ICBA, Caesarea, Israel). Gross milk composition analysis, including fat, protein, and lactose contents, was performed with a MilkoScan FT+ +. SCCs were performed with a Fossomatic FC (Foss Electric, Hilleröd, Denmark). Seventy to eighty percent of female calves are raised on farms for replacement of older cows. Female calves were raised in groups of similar age, artificially inseminated at 13–15 months and transferred to a parturition yard 3 weeks before expected delivery. Post-calving, heifers were relocated to a first-calf heifer group through first lactation. In 2015–2017, all heifers from calving to dry-off entered the study. Every abnormality, with an emphasis on the udder, was recorded and tested by the researcher for infection by CMT on the individual gland, and samples for bacteriology were taken when CMT results were >1 (range: 0, trace, 1, 2, 3). The same procedure was conducted for each of the heifers with SCC above 80,000 cells/mL, following the routine monthly milk-recordings by the ICBA. When diagnoses revealed bacteria and/or only inflammation, two more samples were taken 2–3 days apart. At 70–120 days in milk, at peak lactation, heifers were tested (each quarter) for bacterial infection and milk composition, including whether or not SCC was elevated or abnormalities were present. After verification of each quarter’s condition, composite milk of all non-inflamed glands (no bacteria and SCC <50,000 cells/mL) was sampled during the afternoon milking and sent for analyses of milk composition, SCC, and differential SSC. Milk bacteriology identification was conducted according to the International Dairy Federation [7]. Briefly, 10 µL of each milk sample was inoculated onto blood agar (enriched with 5% washed sheep red blood cells) and MacConkey agar plates (Bacto-Agar,
Difco Laboratory, Becton Dickinson, Le Pont de Claix, France). Plates were incubated at 37 °C and examined for bacterial growth after 18 and 42 h. General data of date of parturition, daily and lactation milk yield, milk composition, number of artificial inseinations (AIs) to pregnancy, length of lactation, medical events, and culling were recorded from the on-line and ICBA Herd Book. Clinical mastitis and some of the sub-clinical mastitis cases were treated with antibiotics.

2.2. Antibodies, Conjugates, and Differential SCC

Somatic cell distributions were tested by flow cytometry (FACs Calibur, Becton-Dickinson, San Jose, CA, USA) as described previously [8]. Briefly, milk samples were analyzed not more than 3 h after collection. The number of somatic cells in each sample was determined by cell counter (Coulter Electronics Ltd., Beds, England) and the volume of milk containing approximately 1 × 10^6 cells was centrifuged (10 min, 230 g, 4 °C), pellets were resuspended in 15 mL phosphate-buffered saline (PBS) and washed by centrifugation (5 min, 200 g, 4 °C). The washed pellets were resuspended in 0.2 mL PBS with a single monoclonal antibody (mAb) or with a pair of mAbs, incubated (1 h at 4 °C), washed (×3) in PBS, and reacted with the conjugated secondary antibodies (30 min, 4 °C). Following incubation, the cells were washed and resuspended in 1 mL PBS. To calculate the percentages of the different leukocytes 10,000 events were read per sample. The commercial mAbs (Kingfisher Biotech Inc. Saint Paul, MN, USA) used for the detection of different leukocytes were: (i) leukocytes, anti–CD18: BAQ30A (IgG-1, RRID:AB_2745264); (ii) lymphocytes, anti–CD4: GC50A1 (IgM, RRID:AB_2848215); (iii) T-cytotoxic lymphocytes, anti–CD8: CACT80C (IgG1, RRID:AB_2848216); (iv) monocytes/macrophages, anti–CD14: CAM36A (IgG1, RRID:AB_2745247); (v) polymorphonuclear (PMN) leukocytes, anti-PMN: CH138A (IgM, RRID:AB_2857323). Paired analyses were performed with anti-CD18 and anti-PMN or anti-CD14 mAbs. All mAbs were species–reactive with bovine cells. Secondary polyclonal antibodies were goat F(ab')2 anti–mouse immunoglobulin IgM (μ-chain specific) conjugated with fluorescein isothiocyanate (FITC) that exhibited minimal cross–reaction to human–, bovine–, and horse–serum proteins (Jackson Immuno Research Lab., West Grove, PA, USA) and affinity isolated goat anti–mouse IgG–1 conjugated with TRI–COLOR (CALTAG Laboratories, Burlingame, CA, USA). The counts of cells presenting these five CD antigens are presented as percentages of the average basal SCC of a healthy gland (44 × 10^3 cells/mL) in Supplementary File S1.

2.3. The Dataset, Haplotype Phasing, and Trait-Association Analysis

Seven traits were analyzed; counts of leukocytes presenting antigens i, ii, iii, iv, v, mean lactation somatic cell score, and infection status. Hairs with roots were pulled from the cows’ tails and submitted for Illumina (San Diego, CA, USA) BovineSNP BeadChip genotyping. Association for bovine genome single-nucleotide polymorphisms (SNPs) was initially determined using PLINK [9], activating the haplotype sliding-window options (–cow –maf 0.05 –hwe 0.0001 –geno 0.1 –hap-window 15 –hap-linear –hap-omnibus). A haplotype near BTA3’s centromere, consisting of 15 SNPs (ARS-BFGL-NGS-30351, ARS-BFGL-NGS-97347, ARS-BFGL-NGS-113990, ARS-BFGL-NGS-8587, ARS-BFGL-NGS-2694, ARS-BFGL-NGS-27948, ARS-BFGL-NGS-6957, BTA-69900-no-rs, ARS-BFGL-NGS-70744, ARS-BFGL-NGS-25382, Hapmap40339-BTA-117016, ARS-BFGL-NGS-87298, ARS-BFGL-NGS-23250, Hapmap49050-BTA-119760, ARS-BFGL-NGS-33062) within nucleotide positions 8,660,956–11,117,776 (build ARS-UCD1.2), was chosen for further analyses. Association for these SNPs was determined using sliding-window PLINK analysis with bootstrapping option (–hap-window 15 –hap-linear –mperm 200000).

2.4. FCGR2 Exon 3 Sequencing and Peak Height Analysis

Template DNA was amplified using PCR primer pair #1 of bovine FCGR2 genes: forward 5'-TCTGAGATTTGGGTGCTGCT-3' and reverse 5'-GCAGATTTTCATCTCCCTTG-3' designed in the second and third introns, respectively. Amplification of these 384-bp amplicons was performed using the Bio-X-ACT™ Long Kit (Bioline Ltd., London, UK) according to the manufacturer’s instructions
and the following conditions: 30 cycles for 40 s at 92 °C, 60 s at 63 °C, and 60 s at 68 °C. In the current genome build (ARS-UCD1.2), these amplicons match BTA3 nucleotide positions 7,942,495–7,942,878 and 8,040,791–8,041,174, within genes FCGR2B and FCGR2A, respectively.

PCR products were separated on agarose gels, excised, purified with AccuPrep® Gel Purification Kit (BioNeer Corp., Seoul, Korea), and then sequenced from both directions using an ABI3730 sequencer and primer pair #1. PCR primers were designed using Primer3 [10]. Nucleotide height or area ratios for the double peaks observed in the ABI trace files were the ratios of the peak values obtained from the poly file generated by Phred base caller [11], activating the write poly file options (phred –d).

2.5. Sequence Read Archive (SRA) Search

SRA BLASTN searches were performed using the dedicated NCBI server [12]. This search was conducted using probes (Table S1) with word size of 32 bp and the Megablast default setting optimized for detection of highly similar sequences. The normalized unit of transcript expression of Reads Per Kilobase of transcript, per Million mapped reads (RPKM) was calculated by 1,000 × (Number of BLAST hits/32) × (10^6/Total mappable reads).

2.6. Deep Sequencing and Analysis of Bovine Genomes

The current reference genome is based on the Hereford beef breed. To find variations between the dairy and beef breeds that may underlie the differences in FCGR2s, DNA was extracted from thawed frozen semen of a single Holstein AI sire (JJ, HOLIDSM000000007424, BioSample: SAMEA5524117) and was deep-sequenced using the next-generation sequencing platform of Illumina HiSeq2000 according to the manufacturer’s paired-end protocol. Average fragment length was 580 bp, and 100-bp sequence reads were obtained from both ends. DNA sample was applied to two lanes, yielding ~30-fold (906,996,192 reads) coverage for this sample. The reference gene sequence was then used as a template for mapping these DNA-Seq reads using GAP5 software [13]. BWA options for this mapping were set to bam bwasw -t 8 -T 60 [14]. The assembled sequence of this sire gene was submitted to ENA under project ID PRJEB38396, ENA accession nos. ERX4137343 (BAM format), and LR798693-6 (annotated gene sequence).

2.7. Statistical Analysis

The genetic correlations among traits or among markers’ substitution effects were estimated as Pearson’s correlation coefficients. These coefficients of correlation were calculated using R package [15] or CORREL function in Excel spreadsheet (Microsoft Corporation, Santa Rosa, CA, USA), respectively. The significance of the correlation coefficient (R), at confidence level of 95% of n observed samples was confirmed by a t-test using the formula: t-value = R × (n−2)^0.5/(1−R^2)^0.5.

3. Results

Aiming for better management of udder health, we recorded immunological, genomic and production parameters of 148 first-calf heifers at the average age of 24 months (Supplementary File S1). In addition to the infection status of each quarter gland, immunological parameters included cytometry counts for five CD markers of milk leukocytes sampled from a healthy gland. These are presented as percentages of the average basal SCC of the healthy gland, which was 44 × 10^3 cells/mL (SE ± 2.5 × 10^3, n = 116). Genomic data consisted of Illumina BeadChip SNP genotypes, and production records were the average values obtained for milk, fat, protein, and SCC at first lactation. According to infection status, we divided the heifers into three groups: uninfected, and clinically and sub-clinically infected; 65 heifers (43.9%) did not have mastitis and mean lactation SCC was 55 × 10^3 cells/mL (SE ± 3 × 10^3, n = 63), and 82 heifers (55.8%) had sub-clinical or clinical infection in one or more glands. With first-lactation average SCCs of 263 ± 35 × 10^3 and of 562 ± 114 × 10^3 in the sub-clinical (n = 58) and clinical (n = 24) groups, respectively, these values significantly differed among the groups.
(p < 0.001). No significant differences were found between the heifer groups with respect to basal CD marker counts in the healthy mammary glands.

3.1. Genome-Wide Association Study (GWAS) of Immunological Traits

Considering adjustments for multiple comparisons and population stratification, we also did not observe any significant association between SNP genotypes and heifer infection status or SCC. However, using PLINK software [9] sliding-window analysis over the bovine genome, we identified an informative haplotype of 15 SNPs in the centromeric end of BTA3 that was strongly associated with the CD18 leukocyte counts (p < 2.3 × 10^{-9}, Figure 1). The identified haplotype consisted of SNPs on the BovineSNP50K BeadChip at positions 9–11 Mb (build ARS-UCD1.2). The PLINK permutation option was employed to further verify the probability of association of the haplotype alleles with CD18 leukocyte counts. For this window, 24 common haplotypes explained >74% of the observed sequence variation (Table 1). Sorting the haplotypes by frequency, the likelihood of association with CD18 leukocyte counts was significant only for presence/absence of the fourth haplotype (#4, F = 5%), which associated with higher counts (β-value 11,600 cells/mL, Table 1). Since this simplified analysis may be confounded by population stratification and multiple comparisons, we applied bootstrapping with 200,000 permutations, which corroborated the significance of this association (EMP2 < 10^{-3}, Table 1). It should be noted that the CD18 counts were highly correlated (R = 0.93) with the PMN cell counts and thus mainly represented the basal level of PMN leukocytes in a healthy mammary gland.

Figure 1. Genome-wide association analysis of basal CD18 leukocyte counts. Omnibus p-values for 15-SNP haplotype association were determined using PLINK. Green arrow points to the most probable association. Black horizontal line denotes the threshold value of significance considering Bonferroni correction for multiple comparisons.

3.2. Analysis of Candidate Genes

Although the gene encoding CD18 (ITGB2) is mapped to the telomeric end of BTA1, the effect on counts of CD18 antigen-presenting cells in milk associated with this locus (T-Stat = 72.3, 15-SNP haplotype between SNPs ARS-BFGL-NGS-4848 and ARS-BFGL-NGS-86458) was much weaker than that recorded for the centromeric end of BTA3 (T-Stat = 90.6, Figure 1). Apparently, in our heifer sample, genetic variation encoded by a BTA3 gene had the major effect on counts of CD18 antigen-presenting cells. To locate this gene, we examined the network of genes that are likely to interact with ITGB2 (Figure 2). This analysis pointed to two genes on BTA3; one of them, FCGR2A (CD32), is a member of the Fcγ receptor (FCGR) gene cluster. This cluster is located at the centromeric position (8 Mb, Figure 2), similar to the SNP haplotype that had the maximal effect on CD18 leukocyte counts. Indeed, in humans, the FCGR gene cluster has long been associated with frequent genetic polymorphism, including copy-number variations (CNVs) that are involved in modulating important immune functions [16]. As we noticed four recorded variations between the reference sequences of the third exons of bovine FCGR2s (NM_001109806; NM_174539), we simultaneously tested whether this genetic variation associates with counts of CD18 antigen-presenting cells, by Sanger sequencing of the third exon (Ex3) of the FCGR2A and 2B genes using PCR primers designed at the adjacent introns. Comparing sequencing chromatograms of high (haplotype #4 carriers) and low CD18-expressing heifers revealed the abundance of the genetic variation encoded by this exon, including a polymorphic 13-bp stretch...
with three nucleotide variations (Figure 3). Sequence chromatograms (n = 15) of the CD32-encoding region showed that heifers with high CD18 counts (n = 4) had a 1:2 copy ratio between antigen genes 2B and 2A, whereas the others exhibited the 1:1 or 0:1 pattern (p < 0.03, Figure 3). Moreover, at the first variable position (G/A, Figure 3), only low-expressing heifers carried the adenine nucleotide, which did not match any reference sequence for FCGR genes, suggesting that this common novel variant may drive low CD18 counts of leukocytes in milk. At this position, most non-carriers of haplotype #4 displayed either 1:1 or 1:2 copy ratios between paralogs (Figure 3).

Table 1. Association of BTA3 haplotype alleles with CD18 leukocyte counts.

| # | HAPLOTYPE | F ¹ | β ¹ | STAT | P ¹ | EMP1 | EMP2 |
|---|-----------|----|----|-----|----|------|------|
| 1 | GGTGAGAACGAAGGG | 0.08 | -1.17 × 10³ | 0.46 | 0.499 | 0.504 | 1.00 |
| 2 | AAAGGAGAAAGGAAGG | 0.07 | -2.17 × 10³ | 1.66 | 0.200 | 0.199 | 1.00 |
| 3 | AGTAGAAAAGAGAAGG | 0.06 | -2.23 × 10³ | 1.17 | 0.291 | 0.282 | 1.00 |
| 4 | GGTGAGGAGGAGAAAG | 0.05 | 1.16 × 10³ | 36.00 | 2.54 × 10⁻⁸ | 5.00 × 10⁻⁶ | 1.00 × 10⁻⁵ |
| 5 | GGTGAGGAGGAGAAAG | 0.04 | 2.11 × 10³ | 0.58 | 0.447 | 0.454 | 1.00 |
| 6 | GGTGAGGAGGAGAAAG | 0.04 | -593 | 0.08 | 0.790 | 0.781 | 1.00 |
| 7 | GGTAAGAAAAGAAAAAG | 0.03 | -4.07 × 10³ | 1.68 | 0.198 | 0.198 | 1.00 |
| 8 | GGTAAGGAGGAGGAAAG | 0.03 | -375 | 0.03 | 0.871 | 0.872 | 1.00 |
| 9 | GGTAGAGGAAAGGAAAG | 0.03 | 4.91 × 10³ | 2.06 | 0.154 | 0.150 | 0.98 |
| 10 | GGTAGGAGGAAAGGAAAG | 0.03 | -550 | 0.04 | 0.849 | 0.850 | 1.00 |
| 11 | AAAAAAGAAGGAAAGG | 0.03 | 1.97 × 10³ | 0.39 | 0.534 | 0.541 | 1.00 |
| 12 | AGTAGGGGAAGGAAAGG | 0.03 | -2.25 × 10³ | 0.34 | 0.559 | 0.572 | 1.00 |
| 13 | AGTAGGAGGAGGAGG | 0.03 | -1.99 × 10³ | 0.33 | 0.564 | 0.576 | 1.00 |
| 14 | GGTGAGAACAAGGAGG | 0.03 | 3.50 × 10³ | 0.63 | 0.429 | 0.449 | 1.00 |
| 15 | GGTGAGAACAAGGAGG | 0.02 | -3.13 × 10³ | 1.33 | 0.251 | 0.253 | 1.00 |
| 16 | ATGGAGAAGAAGGAGG | 0.02 | 5.67 × 10³ | 3.83 | 0.053 | 0.049 | 0.70 |
| 17 | GGTGAGAAAGGAGAAG | 0.02 | -4.01 × 10³ | 1.97 | 0.163 | 0.162 | 0.99 |
| 18 | AGAAGAGGAAAGGAGG | 0.02 | -428 | 0.02 | 0.902 | 0.905 | 1.00 |
| 19 | AGAAGAGGAAAGGAGG | 0.01 | -1.53 × 10³ | 0.20 | 0.638 | 0.668 | 1.00 |
| 20 | GAAAGAGGACAGGAGG | 0.01 | -864 | 0.03 | 0.873 | 0.898 | 1.00 |
| 21 | GGTGAGAAGAAGGAGG | 0.01 | 3.57 × 10³ | 0.87 | 0.353 | 0.365 | 1.00 |
| 22 | GGTGAGAAGAAGGAGG | 0.01 | 480 | 0.01 | 0.929 | 0.944 | 1.00 |
| 23 | GGTGAGGAGGAAAGGAG | 0.01 | -1.45 × 10³ | 0.14 | 0.706 | 0.716 | 1.00 |
| 24 | GGTGAGAAGGAAAGGAG | 0.01 | 3.44 × 10³ | 1.00 | 0.319 | 0.325 | 1.00 |

¹ column definitions: F- frequency in sample, β- regression coefficient, STAT- coefficient t-statistic, P- asymptotic p-value for t-statistic, EMP1- empirical p-value (adaptive), EMP2- corrected empirical p-value. ² Frequency of the haplotype allele was calculated based on 220 BeadChips.

Figure 2. Gene network of the ITGB2 (CD18) gene. Connections between nodes were based on protein–protein interaction data, predicted functional relationships between genes, and gene-pathway data produced by the GeneMANIA system using ITGB2 as the input gene and the Homo sapiens database. On the output of this prediction server, next to the gene nodes, the orthologous bovine autosome numbers and positions were annotated. The nodes of genes with positions coinciding with BTA3 are highlighted in green.
Figure 2. Gene network of the ITGB2 (CD18) gene. Connections between nodes were based on protein–protein interaction data, predicted functional relationships between genes, and gene-pathway data produced by the GeneMANIA system using ITGB2 as the input gene and the Homo sapiens database. On the output of this prediction server, next to the gene nodes, the orthologous bovine autosome numbers and positions were annotated. The nodes of genes with positions coinciding with BTA3 are highlighted in green.

Figure 3. Sequence chromatograms of polymorphic stretch in CD32 third exon of high or low milk CD18-expressing heifers. Tail hairs of high (top) and low (bottom) CD18 leukocyte-count heifers were collected and the extracted DNA was used for simultaneous PCR amplification of the third exons encoding CD32. Forward (F) and reverse (R) orientation chromatograms show three variations (indicated by arrows) in which the ratio of nucleotide heights provides information on the number of paralogs encoding different antigen configurations. In these positions, G-C-A and G-T-G represent the reference sequences of FCGR2A and FCGR2B, transcript IDs: NM_174539 and NM_001109806, respectively.

3.3. CD32 vs. CD18 Gene Expression

Following the indications of genomic variation with functional effect in the bovine FCGR2 genes, we examined whether this variation is transcribed and if current reference sequences of the FCGR2 genes that have been derived from beef cattle also match their repertoire in Holstein cattle. Using 32-bp word BLASTN search (Supplementary Table S1), milk RNA-Seq submissions of 100 Holstein cows deposited in the SRA (project ID PRJNA305942) were searched for the eight possible haplotypes of the polymorphic 13-bp stretch in FCGR2 Ex3 and for the ITGB2 sequence spanning the splice junction of the second and third exons (Supplementary Table S2). Of these probes, probe 9 matching ITGB2 was most frequently expressed, while FCGR2B (probe 4) was frequently detected, and FCGR2A (probe 1) was rarely expressed (Supplementary Table S2). Besides the frequent expression of an unknown FCGR2 variant (probe 7), other probes were not detected or detected only sporadically (Supplementary Table S2). Within this sample, the total read counts varied widely between 60 thousand and 45 million, suggesting that technical factors influenced the quality of the results (Table S2). To further analyze the results, we selected samples with at least 200 probe hits, for which we extended our BLASTN search to RNA-Seq data derived from blood (18 cows, Table 2). The few cows (n = 5) that did not express the unknown FCGR2 variant (probe 7) in milk were the only ones that did not express it in the blood, suggesting that lack of expression emerged from genomic variation. Otherwise, total read numbers and RPKM values showed no correlation between the expression values in milk and in blood. Nevertheless, a significant negative correlation (R = −0.51, p < 0.03) was observed between expression of CD32 and CD18 in milk; and in certain sections, the sample negative R values peaked at over 0.8, as exemplified by sliding-window correlation analysis (Table 2). This correlation supported the
hypothesis that genetic variation in the CD32 gene influences CD18 leukocyte counts in milk but not in blood, which is expected since leukocytes migration into the mammary gland is not likely to have a pronounced effect on the blood reservoir of circulating leukocyte.

Table 2. Comparison of CD32 and CD18 read counts in milk and blood derived from RNA-Seq experiments in Holstein.

|            | Milk | CD32 | CD18 | R  | Blood | CD32 | CD18 |
|------------|------|------|------|----|-------|------|------|
| #1 SRA     | Total| 1    | 4    | 7  | 9     | 1    | 4    |
| 1          | 394  | 27,331,010 | 1 | 16 | 176 | 200 | 27.4 | 201.2 | 270 | 8,584,110 | 5 | 5 | 104 | 36.4 | 378.6
| 2          | 376  | 19,981,694  | 5 | 7 | 7 | 205 | 225 | 31.3 | 320.6 | 226 | 21,522,312 | 0 | 2 | 2 | 173 | 5.8 | 251.2
| 3          | 254  | 18,609,806  | 0 | 3 | 120 | 219 | 232 | 21.8 | 367.7 | 250 | 1,177,238  | 0 | 0 | 16 | 0.0 | 424.7
| 4          | 367  | 16,301,126  | 3 | 2 | 10 | 220 | 235 | 28.8 | 421.8 | 218 | 22,950,370 | 0 | 13 | 17 | 192 | 40.8 | 261.4
| 5          | 368  | 20,676,820  | 0 | 19 | 7 | 209 | 235 | 38.9 | 312.8 | 170 | 14,416,080 | 0 | 17 | 16 | 164 | 71.5 | 355.5
| 6          | 340  | 21,342,290  | 10 | 54 | 0 | 173 | 237 | 93.7 | 253.3 | 235 | 13,338,926 | 0 | 26 | 0 | 144 | 60.9 | 337.4
| 7          | 341  | 25,687,198  | 4 | 54 | 0 | 173 | 237 | 93.7 | 253.3 | 235 | 13,338,926 | 0 | 26 | 0 | 144 | 60.9 | 337.4
| 8          | 210  | 14,517,104  | 2 | 6 | 9 | 235 | 252 | 36.6 | 505.9 | 257 | 9,046,998  | 0 | 0 | 7 | 103 | 27.6 | 355.8
| 9          | 361  | 23,695,512  | 0 | 22 | 9 | 247 | 279 | 42.2 | 325.7 | 207 | 13,026,116 | 0 | 15 | 8 | 274 | 55.2 | 657.2
| 10         | 232  | 17,678,968  | 0 | 3 | 306 | 309 | 5.3 | 540.9 | −0.55 | 168 | 23,333,558 | 0 | 2 | 34 | 122 | 48.2 | 163.4
| 11         | 188  | 40,398,796  | 3 | 42 | 8 | 277 | 330 | 41.0 | 214.3 | −0.65 | 214 | 23,555,914 | 1 | 26 | 20 | 223 | 63.7 | 295.8
| 12         | 345  | 26,493,242  | 2 | 40 | 0 | 299 | 341 | 49.5 | 352.7 | −0.67 | 193 | 23,755,222 | 0 | 51 | 0 | 307 | 67.1 | 403.9
| 13         | 318  | 41,897,140  | 0 | 177 | 103 | 68 | 348 | 208.8 | 50.7 | −0.81 | 229 | 6,410,488 | 0 | 9 | 5 | 53.6 | 287.6
| 14         | 321  | 45,013,510  | 15 | 32 | 32 | 362 | 441 | 54.8 | 251.3 | −0.79 | 230 | 21,228,336 | 1 | 19 | 19 | 303 | 504.9 | 446.0
| 15         | 365  | 18,179,706  | 0 | 7 | 3 | 439 | 449 | 17.2 | 754.6 | −0.73 | 258 | 11,903,986 | 0 | 9 | 4 | 157 | 33.9 | 409.4
| 16         | 316  | 30,449,572  | 0 | 24 | 15 | 410 | 449 | 40.0 | 420.8 | −0.72 | 261 | 3,872,402 | 0 | 7 | 1 | 61 | 64.6 | 492.3
| 17         | 243  | 28,918,832  | 8 | 22 | 12 | 443 | 485 | 45.4 | 478.7 | −0.74 | 217 | 18,766,692 | 3 | 11 | 214 | 55.0 | 356.3
| 18         | 349  | 30,480,908  | 10 | 88 | 0 | 541 | 639 | 100.5 | 554.7 | −0.61 | 273 | 8,150,358 | 0 | 12 | 0 | 119 | 46.0 | 456.3

1 Each line presents an individual cow, omitting the SRX2235 prefix from its matching SRA IDs. 2 Table is sorted by the sum of counts of reads for CD32 transcript-matching probes 1–8 and, probe 9 matching CD18. 3 Correlation was calculated between the preceding CD32 and CD18 RPKM records over a sliding window of 10 lines.

3.4. Computerized Cloning of FCGR Genes of an Influential Israeli Holstein Sire

Since the sequence of the unknown FCGR2 variant (probe 7) matched the novel variant detected by Sanger sequencing (adenine nucleotide in the position of the first variation, Figure 3), we chose to further characterize this transcript, which was abundant in Holstein cattle and thus may reflect adaptation to selection pressure toward low leukocyte counts in Holstein milk induced by the breeding program. We assembled expressed sequence tags (ESTs) of four clones (EST IDs: FE019948-9, FE020120-1, EE937361-2, and BE845572), realizing that this transcript was similar to FCGR2B (GenBank ID: NM_174539) except for a missense variation resulting in a non-conservative serine-to-glycine substitution in the predicted extracellular structure of the immunoglobulin-like C-2 type, which is encoded by Ex3 (Figure 4).

Using the 32-bp word BLASTN search, we meta-analyzed genomic DNA-Seq submissions (SRA, project ID PRJNA343262) of 20 dairy (Holstein) and 20 beef (Angus, Hereford, Charolaise, Limousine, Simmental, four individuals each) cattle. The number of hits for the eight possible haplotypes of the polymorphic 13-bp stretch in FCGR2 Ex3 was compared to those for three control genes (HOXB7, HOXA7, SRY), for which CNV is assumed to be rare (Supplementary Table S2). The ratio of hits for probe 7 and total hits for FCGR2 was indeed significantly higher for Holsteins (24%, p < 0.01) than for beef cattle (9%). Nevertheless, while this probe was not detected in the Hereford and Limousine individuals, it was common in the others (e.g., Simmental, 21%). FCGR2 copy numbers based on correction for coverage and average detection rate of the control gene probes fluctuated between two and nine (Supplementary Table S2).

To better study the FCGR2 gene sequence and copy number in dairy cattle, we analyzed a representative Holstein sire (JJ, HOLISRM000000007424), applying deep sequencing to the genome. At the end of 2018, this sire was recorded in the top 20 sires for total net merit, with more than 10,000 daughters. Being a descendant of the popular US bull O–Bee Manfred Justice (HOU SA000122358313), this sire represents an influential bloodline of Holstein cattle. Directed assembly was hampered by the presence of multiple similar gene copies. Focusing on the most polymorphic third exon, at least four
distinct alleles were detected (Figure 4). Following our eight probe sequences in FCGR2 Ex3, we named these alleles/variants using the prefix "V" (i.e., V1a, V1b, V4a, V4b, V4c, and V7). For each of the six contigs, relative copy number was estimated based on the number of reads that spanned the central variation in the polymorphic 13-bp stretch in FCGR2 Ex3. The count in the V7 contig (n = 28) was similar to the combined count of the other contigs (n = 25, consisting of V1a, V1b, V4a, V4b, and V4c with 7, 3, 2, 11, and 2 reads, respectively). This suggested that at least five copies, designated V7a—V7e, were present in this sire genome. To further support this hypothesis, we performed peak-height analysis of Sanger chromatograms obtained for the Ex3 amplicon. Indeed, the adenine residue typical of V7 displayed equal peak heights compared to the guanine base of the other variants (Figure 4).

In the other four polymorphic positions, peak heights also well fitted the model described in Figure 4, in which 10 allele copies of Ex3 were co-sequenced, resulting in the observed superimposed pattern. Thus, the examined sire had a CNV with at least five FCGR2 genes, compared to the two genes in the bovine reference genome. Moreover, the number of V7 allele copies (n = 5) indicated that it could have arisen from an independent paralog, which we named FCGR2C (ENA accession no. LR798694).

Although the transcript of the FCGR2C variant was almost identical to that of FCGR2B, the introns in the assembled FCGR2C contig were more similar to those of FCGR2A (Experiment accession no. ERX4137343).

**Figure 4.** Sequence chromatograms of CD32 third exon of a representative Holstein sire compared to an allele model based on DNA-Seq. DNA extracted from semen was used for simultaneous PCR amplification of the third exons encoding CD32. Chromatograms show five variations in which the ratio of nucleotide heights provided information on the number of encoding paralogs. Below the chromatogram, dots indicate similarity to the consensus sequence of 10 allele variants predicted by the assembled sequences and counts of this sire’s DNA-Seq reads. Putative amino acid translation is given below the consensus sequence, in which codons are annotated by alternating font and background color. Nucleotide and amino acid variations are highlighted in yellow. Arrowheads point to the two cysteines predicted to form the disulfide bond that stabilizes an extracellular structure of the immunoglobulin-like C-2 type.

### 4. Discussion

SCC in cows’ milk is a useful indicator of milk quality and of mastitis, the costliest disease in dairy production. Understanding the genetic architecture of immune traits is essential to combating mastitis, and SCC is the most widely used trait in GWASs that aim to examine this architecture. However, in a
A typical mammalian genome, about one-tenth of the genes with recorded GO annotation for biological process are of immune system processes, e.g., in humans, 3,263/34,673 [17]. This abundance of inflammatory and immunological genes hinders GWASs, which are likely to falsely map “ghost” QTLs that arise due to the superposition of additive and epistatic polygenic effects near chromosomal regions that are rich in immune system genes. Albeit at the cost of a loss in power, statistical methods can control for another type of false-positive QTL, which occurs in structured populations; the “ghost” QTL problem is further complicated by such loci that are in linkage disequilibrium with a causal variant [18]. These complications may explain why in general, the overlap in genome-mapping results on QTLs affecting clinical mastitis and somatic cell score is low, even in Nordic countries where the breeding system is best at recording mastitis [19]. Knowledge of the proportion of somatic cell types in milk, rather than just the overall SCC, provides valuable information on the inflammatory status of the udder, and differential SCC has heritability (0.08) that is double that of traditional SCCs [20]. Indeed, when controlling for multiple comparisons and population stratification, the association between SNP genotypes and heifer infection status or conventional SCC was insignificant for our sample, whereas the centromeric end of BTA3 was highly associated with the differential SCC obtained by flow cytometry for leukocytes presenting CD18. To the best of our knowledge, this is the first report of a GWAS of differential SCCs based on flow cytometry of CD antigens in milk. It should be noted that CD18 has been previously associated with leukocyte performance as a result of bovine leukocyte adhesion deficiency (BLAD), which is characterized by a lack of CD11/CD18 adhesion molecules on the leukocyte surface. This syndrome is not likely to affect our sample’s heifers as they were sired by bulls that are screened against BLAD.

Searching for a CD18-effector gene in the centromeric end of BTA3, we noticed an FCGR gene cluster, including FCGR2A (CD32), which belongs to the network of genes interacting with ITGB2 (CD18) [21]. Known for CNV of FCGR genes, the orthologous cluster on HSA1 likely evolved through tandem duplication of type II and type III low-affinity FCGRs, in which a heat-shock protein family A gene (HSP70) is embedded [22]. Corroborating this synteny, bovine HSPA6 is positioned between FCGR2A and FCGR3A in the current Bos taurus genome build (ARS-UCD1). Further support for the candidacy of FCGR2s as genes involved in the regulation of CD18 SCC was provided by the Sanger-sequence analysis of amplicons derived from their polymorphic third exons. This co-amplification yielded chromatogram patterns that differed between cows with high and low CD18 SCCs and thus significantly linked CD18 counts to nucleotide variation in CD32 (∼0.03). A polymorphic 13-bp stretch in FCGR2 Ex3 with three adjacent SNPs proved useful for tagging FCGR2 variants, and out of the eight possible nucleotide combinations, using BLASTN searches of public nucleotide databases, we observed three variants (tags v1, v4, and v7). The bovine reference genome is based on Hereford beef cattle and lacks essential comparative information for other breeds [23] and for FCGR2s, the annotation includes FCGR2A (Gene ID: 782652) and FCGR2B (Gene ID: 282229), which correspond to v1 and v4, respectively. This annotation is compatible with the absence of the v7 variant in Hereford individuals and the indication of FCGR2 CNV based on comparisons of read counts between FCGR2 tags and control genes (Supplementary Table S3). This method has previously been used to discover CNV in bovine FCGR2B (chr3:8,648,846–8,665,349, Table_S7_CNVR_RefSeq) [24].

Calculating peak–height ratios between SNP alleles in sequencing chromatograms is another practical method for estimating their copy number proportions, especially when similar ratios are obtained by sequencing from the forward and reverse orientations [25,26]. To better characterize v7, which was common in Holsteins (75% of individuals, Supplementary Table S3), we compared the sequence assembly of FCGR2 Ex3 based on next-generation sequencing to the Sanger-sequencing chromatograms of this exon from an influential Israeli AI sire. Results of this analysis were compatible with a model of five paralogous FCGR2s and expansion of the v7 paralog, which was designated FCGR2C. This polymorphic nature and the recent burst of duplication and divergence of the bovine FCGR2C are typical of immune response genes under selective pressure [27]. Such pressure can be exerted by Holstein breeding programs that select for lower SCC [28].
Seeking further confirmation of the hypothesized CD18–CD32 gene interplay, we meta-analyzed RNA-Seq data, comparing the expression levels of these antigens. This revealed a significant negative correlation ($R = -0.51$) between the expression of CD32 and CD18 in milk. Thus, it is possible that extra gene copies of CD32 enhance its expression and induce immunomodulation towards low CD18 SCC. In uninfected glands, leukocytes are incorrectly thought to represent the majority of the somatic cells, with a few cells that are shed from the epithelium [28]. Yet, in these glands, epithelial cells accounted for ~50% of the SCC in cows [29,30]. Therefore, using flow cytometry to detect surface CD18 may be crucial for analyses of leukocytes in basal SCCs, as CD18 is a general marker of these cells and is essential for their passage into milk [31]. Controversy exists about whether low milk cell counts increase susceptibility to mastitis [28]. We observed no correlation between mastitis and basal SCC, supporting the conclusion of [28] that a low SCC resulting from selection has no negative effect on susceptibility to mastitis. Considering the current index for selection, this scenario clearly favors individuals with a genetic architecture that supports low basal SCC combined with efficient reactivity of the healthy glands to infection. Thus, the increased copy number of FCGR2s and abundance of FCGR2C in dairy cattle probably reflect adaptation to the selection pressure towards low SCC in Holstein milk.

**Supplementary Materials:** The following are available online at http://www.mdpi.com/2073-4425/11/8/952/s1, File S1: Dataset (includes ID numbers of cow, sire, dam, paternal and maternal grand sires; liters of milk yield on day of testing; mean SCC recorded for the relevant lactation; health group; bacteriology of the glands; percentages of cells with surface antigens CD18, PMN, CD4, CD8, CD14, BTA3; SNP alleles of BTA3 haplotypes), Table S1: Nucleotide sequences (32 bp probes) used for Megablast search (fasta format), Table S2: Comparison of number of RNA-probe hits and expression levels between FCGR2s (CD32) and CD18, in Holstein milk, Table S3: Comparison of number of genomic-probe hits and gene copies between FCGR2s (CD32) and control genes.

**Author Contributions:** S.E.B designed and analyzed the flow cytometry experiments, which were performed by O.K.; G.L. performed the immunological diagnosis; A.S. performed Sanger-sequencing analyses; S.J. managed the cow shed and production records; E.S. conducted the bioinformatics and GWAS analyses. G.L. and E.S. conceived and designed the experiments and drafted the manuscript; all authors wrote, read, and approved the article.

**Funding:** This research was supported by a grant from the research fund of the Israel Dairy Board and grant #20-11-0009 from The Chief Scientist of the Israeli Ministry of Agriculture.

**Acknowledgments:** The authors are grateful to the teams of the dairy herds for allowing us to conduct this study and for their technical assistance and care of the animals. The authors thank Yaniv Lavon of ICBA for helpful comments.

**Conflicts of Interest:** The authors declare no conflict of interest.

**References**

1. Halasa, T.; Huijps, K.; Osteras, O.; Hogeveen, H. Economic effects of bovine mastitis and mastitis management: A review. *Vet. Q.* 2007, 29, 18–31. [CrossRef]
2. Abebe, R.; Hatiya, H.; Abera, M.; Megersa, B.; Asmare, K. Bovine mastitis: Prevalence, risk factors and isolation of *Staphylococcus aureus* in dairy herds at Hawassa milk shed, South Ethiopia. *BMC Vet. Res.* 2016, 12, 270. [CrossRef] [PubMed]
3. Leitner, G.; Blum, S.E.; Krifuks, O.; Edery, N.; Merin, U. Correlation between Milk Bacteriology, Cytology and Mammary Tissue Histology in Cows: Cure from the Pathogen or Recovery from the Inflammation. *Pathogens* 2020, 9, 364. [CrossRef] [PubMed]
4. Leitner, G.; Shoshani, E.; Krifuks, O.; Chaffer, M.; Saran, A. Milk leucocyte population patterns in bovine udder infection of different aetiology. *J. Vet. Med. B* 2000, 47, 581–589. [CrossRef] [PubMed]
5. Paape, M.J.; Shafer-Weaver, K.; Capuco, A.V.; Van Oostveldt, K.; Burvenich, C. Immune Surveillance of Mammary Tissue by Phagocytic Cells. In *Biology of the Mammary Gland*; Springer US: Boston, MA, USA, 2002; pp. 259–277.
6. Seroussi, E.; Blum, S.E.; Krifuks, O.; Lavon, Y.; Leitner, G. Application of pancreatic phospholipase A2 for treatment of bovine mastitis. *PLoS ONE* 2018, 13, e0203132. [CrossRef]
7. Oliver, S.P.; Gonzalez, R.N.; Hogan, J.S.; Jayarao, B.M.; Owens, W.E. *Microbiological Procedures for the Diagnosis of Bovine Udder Infection and Determination of Milk Quality*, 4th ed.; The National Mastitis Council, Inc.: Verona, WI, USA, 2004.

8. Leitner, G.; Eliligulashvily, R.; Krifucks, O.; Perl, S.; Saran, A. Immune cell differentiation in mammary gland tissues and milk of cows chronically infected with *Staphylococcus aureus*. *J. Vet. Med. Ser. B* 2003, 50, 45–52. [CrossRef]  

9. Purcell, S.; Neale, B.; Todd-Brown, K.; Thomas, L.; Ferreira, M.A.R.; Bender, D.; Maller, J.; Sklar, P.; de Bakker, P.I.W.; Daly, M.J.; et al. PLINK: A tool set for whole-genome association and population-based linkage analyses. *Am. J. Hum. Genet.* 2007, 81, 559–575. [CrossRef]  

10. Bonfield, J.K.; Whitwham, A. Gap5—Editing the billion fragment sequence assembly. *Bioinformatics* 2010, 26, 1699–1703. [CrossRef]  

11. Ewing, B.; Hillier, L.; Wendl, M.C.; Green, P. Base-calling of automated sequencer traces using Phred. I. Accuracy assessment. *Genome Res.* 1998, 8, 175–185. [CrossRef]  

12. NCBI. Sequence Read Archive Nucleotide BLAST. Available online: https://blast.ncbi.nlm.nih.gov/Blast.cgi?PROGRAM=blastn&PAGE_TYPE=BlastSearch&PROG_DEFAULTS=on&BLAST_SPEC=SRA (accessed on 30 June 2020).

13. Bonfield, J.K.; Whitwham, A. Gap5—Editing the billion fragment sequence assembly. *Bioinformatics* 2010, 26, 1699–1703. [CrossRef]  

14. Li, H.; Durbin, R. Fast and accurate long-read alignment with Burrows-Wheeler transform. *Bioinformatics* 2010, 26, 589–595. [CrossRef] [PubMed]  

15. R-Core-Team. R: A Language and Environment for Statistical Computing. Available online: http://www.R-project.org/ (accessed on 30 June 2020).

16. Fanciulli, M.; Vyse, T.J.; Aitman, T.J. Copy number variation of Fcγ receptor genes and disease predisposition. *Cytogenet. Genome Res.* 2008, 123, 161–168. [CrossRef] [PubMed]  

17. Carbon, S.; Ireland, A.; Mungall, C.J.; Shu, S.; Marshall, B.; Lewis, S. AmiGO: Online access to ontology and annotation data. *Bioinformatics* 2009, 25, 288–289. [CrossRef] [PubMed]  

18. Mott, R.; Flint, J. Dissecting quantitative traits in mice. *Annu. Rev. Genom. Hum. Genet.* 2013, 14, 421–439. [CrossRef] [PubMed]  

19. Lund, M.S.; Guildbrandtsen, B.; Buitenhuis, A.J.; Thomsen, B.; Bendixen, C. Detection of quantitative trait loci in Danish Holstein cattle affecting clinical mastitis, somatic cell score, udder conformation traits, and assessment of associated effects on milk yield. *J. Dairy Sci.* 2008, 91, 4028–4036. [CrossRef] [PubMed]  

20. Bobbo, T.; Penasa, M.; Cassandro, M. Short communication: Genetic aspects of milk differential somatic cell count in Holstein cows: A preliminary analysis. *J. Dairy Sci.* 2019, 102, 4275–4279. [CrossRef]  

21. Montojo, J.; Zuberi, K.; Rodriguez, H.; Bader, G.D.; Morris, Q. GeneMANIA: Fast gene network construction and function prediction for Cytoscape. *F1000Res* 2014, 3, 153. [CrossRef]  

22. Breunis, W.B.; van Mirre, E.; Geissler, J.; Laddach, N.; Wolbink, G.; van der Schoot, E.; de Haas, M.; de Boer, M.; Roos, D.; Kuipers, T.W. Copy number variation at the FCGR locus includes FCGR3A, FCGR2C and FCGR3B but not FCGR2A and FCGR2B. *Hum. Mutat.* 2009, 30, E640–E650. [CrossRef]  

23. Kohs, S.; Reimann, E.; Lilleoja, R.; Lattekivi, F.; Salumets, A.; Reemann, P.; Jaakma, U. Sequencing and annotated analysis of full genome of Holstein breed bull. *Mamm. Genome* 2014, 25, 363–373. [CrossRef]  

24. Bickhart, D.M.; Hou, Y.L.; Schroeder, S.G.; Alkan, C.; Cardone, M.F.; Matukumalli, L.K.; Song, J.Z.; Schnabe, R.D.; Ventura, M.; Taylor, J.F.; et al. Copy number variation of individual cattle genomes using next-generation sequencing. *Genome Res.* 2012, 22, 778–790. [CrossRef]  

25. Seroussi, E.; Klimpus, S.; Silanikove, M.; Krifucks, O.; Shapiro, F.; Gertler, A.; Leitner, G. Nonbactericidal secreted phospholipase A2s are potential anti-inflammatory factors in the mammary gland. *Immunogenetics* 2013, 65, 861–871. [CrossRef] [PubMed]  

26. Shirak, A.; Seroussi, U.; Gootwine, E.; Seroussi, E. Sequence motifs capable of forming DNA stem-loop structures act as a replication diode. *FEBS Open Bio* 2017, 7, 944–952. [CrossRef] [PubMed]  

27. Golik, M.; Cohen-Zinder, M.; Loor, J.J.; Drackley, J.K.; Band, M.R.; Lewin, H.A.; Weller, J.I.; Ron, M.; Seroussi, E. Accelerated expansion of group IID-like phospholipase A2 genes in Bos taurus. *Genomics* 2006, 87, 527–533. [CrossRef] [PubMed]  

28. Rainard, P.; Foucras, G.; Boichard, D.; Rupp, R. Invited review: Low milk somatic cell count and susceptibility to mastitis. *J. Dairy Sci.* 2018, 101, 6703–6714. [CrossRef] [PubMed]
29. Kelly, A.L.; Leitner, G.; Merin, U. Milk quality and udder health. Test methods and standards. In Encyclopedia of Dairy Sciences, 2nd ed.; Fuquay, J.W., Fox, P.F., McSweeney, P.L.H., Eds.; Academic Press: San Diego, CA, USA, 2011; pp. 894–901.

30. Leitner, G.; Merin, U.; Krifucks, O.; Blum, S.; Rivas, A.L.; Silanikove, N. Effects of intra-mammary bacterial infection with coagulase negative Staphylococci and stage of lactation on shedding of epithelial cells and infiltration of leukocytes into milk: Comparison among cows, goats and sheep. Vet. Immunol. Immunopathol. 2012, 147, 202–210. [CrossRef] [PubMed]

31. Smits, E.; Burvenich, C.; Guidry, A.J.; Massart-Leen, A. Adhesion receptor CD11b/CD18 contributes to neutrophil diapedesis across the bovine blood-milk barrier. Vet. Immunol. Immunopathol. 2000, 73, 255–265. [CrossRef]

© 2020 by the authors. Licensee MDPI, Basel, Switzerland. This article is an open access article distributed under the terms and conditions of the Creative Commons Attribution (CC BY) license (http://creativecommons.org/licenses/by/4.0/).