Validation of PDE9A Gene Identified in GWAS Showing Strong Association with Milk Production Traits in Chinese Holstein

Shao-Hua Yang †, Xiao-Jun Bi †, Yan Xie, Cong Li, Sheng-Li Zhang, Qin Zhang and Dong-Xiao Sun *

Received: 17 June 2015; Accepted: 18 August 2015; Published: 5 November 2015

College of Animal Science and Technology, Key Laboratory of Animal Genetics and Breeding of Ministry of Agriculture, National Engineering Laboratory of Animal Breeding, China Agricultural University, Beijing 100193, China; yangshaohua309@126.com (S.-H.Y.); yaning812@126.com (X.-J.B.); xiey.300822@aliyun.com (Y.X.); li10020902@163.com (C.L.); zhang62733697@163.com (S.-L.Z.); qzhang@cau.edu.cn (Q.Z.)

* Correspondence: sundx@cau.edu.cn; Tel./Fax: +86-10-6273-4653
† These authors contributed equally to this work.

Abstract: Phosphodiesterase9A (PDE9A) is a cyclic guanosine monophosphate (cGMP)-specific enzyme widely expressed among the tissues, which is important in activating cGMP-dependent signaling pathways. In our previous genome-wide association study, a single nucleotide polymorphism (SNP) (BTA-55340-no-rs) located in the intron 14 of PDE9A, was found to be significantly associated with protein yield. In addition, we found that PDE9A was highly expressed in mammary gland by analyzing its mRNA expression in different tissues. The objectives of this study were to identify genetic polymorphisms of PDE9A and to determine the effects of these variants on milk production traits in dairy cattle. DNA sequencing identified 11 single nucleotide polymorphisms (SNPs) and six SNPs in 5′ regulatory region were genotyped to test for the subsequent association analyses. After Bonferroni correction for multiple testing, all these identified SNPs were statistically significant for one or more milk production traits (p < 0.0001~0.0077). Interestingly, haplotype-based association analysis revealed similar effects on milk production traits (p < 0.01). In follow-up RNA expression analyses, two SNPs (c.-1376 G>A, c.-724 A>G) were involved in the regulation of gene expression. Consequently, our findings provide confirmatory evidences for associations of PDE9A variants with milk production traits and these identified SNPs may serve as genetic markers to accelerate Chinese Holstein breeding program.

Keywords: dairy cattle; PDE9A; milk production traits; association analysis; GWAS; gene expression

1. Introduction

Milk production traits are the most economically important traits controlled by numerous genes and environmental factors in dairy cattle. Therefore, an improvement in milk production traits continues to be the most profitable breeding objective. Mutations can alter breeding values of economical traits in dairy cattle [1]. New molecular techniques focused on genome analysis have made it feasible to screen for mutations associated with complex traits by genome-wide association study (GWAS) [2]. Compared with traditional QTL (quantitative trait loci) mapping strategy, GWAS shows obvious advantages both in the power to detect harboring variants and in simplifying the discovery of causal variants [3]. Thus, GWAS have been widely recognized as an important strategy to explore genes associated with complex traits in many species. Using Illumina 50K chip, we identified 105 genome-wide significant SNPs associated with milk production traits in Chinese Holstein...
Holstein population. Among these SNPs, a SNP, BTA-55340-no-rs (\(p = 9.72 \times 10^{-7}, n = 1815\)), located in the intron 14 of the phosphodiesterases 9A (PDE9A) gene was highly significant with effects on protein yield [4]. The PDE9A gene is located on bovine BTA 1, which included a number of identified QTLs for milk production traits [5,6]. These findings strongly suggest that PDE9A may be one of the most promising novel candidates for milk production traits.

PDEs, a large family of enzymes specifically hydrolyzing the second messenger cAMP and/or cGMP, are the only cellular mechanism for degrading cAMP and cGMP [7] and thus play a critical role in regulating the intracellular levels of these second messengers and subsequently functional responses of cells [8–11]. The PDE9A gene encodes a cGMP-specific high-affinity PDE that seems to be widely expressed among tissues or organs in humans and mice [12,13]. The bovine PDE9A gene has 19 exons spanning 110 kb. According to the significant evidences in our GWAS as well as comparative genomics, we therefore assumed that the PDE9A gene could be a positional and functional candidate gene for milk production traits in Chinese Holstein population.

As a novel candidate for milk composition traits detected by GWAS, however, the bovine PDE9A gene has not been reported in relation with milk production traits. Therefore, the purpose of the present study was to search for potential casual genetic variants associated with milk production traits in an independent dairy cattle population. Furthermore, tissue expression pattern and the role of regulatory SNPs on their mRNA expression were also investigated. The results showed that the two identified variants in 5' regulatory region of PDE9A may be important genetic factors involved in milk production ability and can be used in marker-assisted breeding on the basis of further validation.

2. Results and Discussion

2.1. Molecular Evolutionary Analysis of the Bovine PDE9A Gene

Based on protein sequence of bovine PDE9A (XP_005202092.1), we obtained the following similarities by comparing of the PDE9A amino acid sequence with seven other animal species (Table 1): Homo sapiens (85.39%), Rattus norvegicus (90.52%), Mus musculus (91.33%), Gallus gallus (81.64%), Danio rerio (77.81%), Capra hircus (98.81%) and Ovis aries (98.11%). The results suggested that the PDE9A gene was extremely conserved within mammalian species.

| Species       | GenBank Accession | Similarity |
|---------------|-------------------|------------|
| Capra hircus  | XP_005675698.1    | 98.81%     |
| Ovis aries    | XP_004003956.1    | 98.11%     |
| Mus musculus  | NP_001157220.1    | 91.33%     |
| Rattus norvegicus | XP_008771017.1 | 90.52%     |
| Homo sapiens  | NP_001001567.1    | 85.39%     |
| Gallus gallus | XP_003640550.1    | 81.64%     |
| Danio rerio   | XP_009302929.1    | 77.81%     |

A phylogenetic tree was constructed using the MEGA 4.0.2 software to better understand the potential evolutorial process. According to the phylogenetic tree (Figure 1), the bovine PDE9A was found to be phylogenetically closest to Capra hircus and Ovis aries PDE9A. Meanwhile, it was closely associated with the mouse and rat PDE9A, with the Homo sapiens forming a separate group, while the non-mammalian species formed an even more distant group.
2.2. Expression Analysis of the Bovine PDE9A Gene

Tissue distribution analysis of bovine PDE9A was determined by quantitative real-time PCR and the relative expression results were normalized by internal GAPDH expression. As shown in Figure 2, PDE9A was widely expressed in all these detected tissues, with higher expression level in mammary gland, moderately expressed in small intestine, uterus and ovary, and only slightly expressed in liver, kidney, heart and gluteus. The highest mRNA expression level of PDE9A gene in mammary gland indicated that it was most likely related to the formation of milk related traits. As observed in previous investigations, some identified functional genes for milk production traits showed the same specially high expression patterns in lactating mammary tissue of mammals, that is, DGAT1 [14], ABCG2 [15] and SCD1 [16].

2.3. SNP Identification and Selection

Sequence analysis revealed that a total of 11 SNPs were found in the PDE9A gene using the pooled DNA of the eight unrelated bulls. Of them, six SNPs were in the 5′ regulatory region, two in intron14 and three in 3′ regulatory region. Considering the positions and functions of these polymorphisms, six SNPs in the 5′ regulatory region were finally chosen and genotyped for further association analysis. These selected SNPs were all in Hardy-Weinberg equilibrium (chi-square test, \( p > 0.05 \)), and their genotypic and allelic frequencies are shown in Table 2.
Table 2. Genotypes and allelic frequencies of the six SNPs in the 5' regulatory region of PDE9A.

| SNP ID | GenBank No. | Allele Substitution | Genotypes | Genotypic Frequencies | Alleles | Allelic Frequencies |
|--------|-------------|---------------------|-----------|-----------------------|---------|-------------------|
| c.-2012 T>C | rs42140305  | T>C                  | TT        | 0.40 T                | 0.64    |
|         |             |                     | CT        | 0.48                  |
|         |             |                     | CC        | 0.12 C                | 0.36    |
| c.-2005 A>G | rs381951806 | A>G                  | AA        | 0.56 A                | 0.74    |
|         |             |                     | AG        | 0.37                  |
|         |             |                     | GG        | 0.07 G                | 0.26    |
| c.-1762 T>C | rs136803034 | T>C                  | TT        | 0.37 T                | 0.61    |
|         |             |                     | CT        | 0.49                  |
|         |             |                     | CC        | 0.14 C                | 0.39    |
| c.-1621 T>G | rs133033048 | T>G                  | GT        | 0.38                  |
|         |             |                     | GG        | 0.10 G                | 0.28    |
| c.-1376 G>A | rs210080144 | G>A                  | GG        | 0.65 G                | 0.81    |
|         |             |                     | AG        | 0.32                  |
|         |             |                     | AA        | 0.03 A                | 0.19    |
| c.-724 A>G | rs135748250 | A>G                  | AA        | 0.38 A                | 0.64    |
|         |             |                     | AG        | 0.52                  |
|         |             |                     | GG        | 0.10 G                | 0.36    |

2.4. Association Analyses

2.4.1. Single Locus-Based Regression Analyses

Associations between the six SNPs and estimated breeding values (EBVs) of five milk production traits are shown in Table 3. Here, all these SNPs were statistically significant with protein yield (PY) (p < 0.0001~0.0077) and milk yield (MY) (p < 0.0001~0.0015; except for c.-1621 T>G). Meanwhile, the c.-1762 T>C, c.-1376 G>A and c.-724 A>G were strongly associated with fat yield (FY) (p = 0.0016~0.0009). After Bonferroni correction for multiple testing, these identified associations remained significant. Therefore, our results provide a valuable primary evidence for identifying candidate genes in dairy cattle.

Similarly, additive and allele substitution effects were observed for the three associated traits at all six loci as well (p < 0.05 ~ p < 0.01). The results of additive effects were mostly in consistent with those of allele substitution effects, which provided strong evidence for accuracy of additive genetic variance based on phenotypic individual differences. As for the dominant effect, only c.-1621 T>G reached significance on PY (p < 0.01), indicating that the genetic value for GT is not exactly the average of the genetic value of GG and TT. The estimated effects of the six SNPs on five traits are shown in Table 4.

Table 3. Associations of SNPs with EBVs for five milk production traits (least squares mean ± standard error).

| SNP    | Genotype | MY   | FY   | FP   | PY   | PP   |
|--------|----------|------|------|------|------|------|
| c.-2012 T>C | CC (59)  | 531.94 ± 102.58 A | 11.44 ± 4.38 | -0.05 ± 0.046 | 15.89 ± 3.06 A | -0.006 ± 0.015 |
|         | CT (239) | 363.53 ± 75.15 A  | 3.62 ± 3.39  | -0.07 ± 0.035  | 10.82 ± 2.37 A,B | -0.002 ± 0.010 |
|         | TT (198) | 192.65 ± 78.00 B  | 0.91 ± 3.49  | -0.04 ± 0.036  | 7.10 ± 2.44 B    | 0.012 ± 0.011  |
| p Value | 0.0002 ** | 0.0115 | 0.5315 | 0.0009 ** | 0.2203 |
| c.-2005 A>G | AA (275) | 242.80 ± 74.98 A  | 6.24 ± 3.36  | 0.01 ± 0.023   | 9.72 ± 2.35 A    | 0.03 ± 0.005  |
|         | AG (183) | 361.60 ± 79.38 A,B | 6.77 ± 3.50  | 0.01 ± 0.030   | 12.42 ± 2.45 A,b | 0.003 ± 0.006  |
|         | GG (38)  | 596.23 ± 118.54 B | 11.44 ± 5.11 | 0.01 ± 0.034   | 18.84 ± 3.58 b   | -0.014 ± 0.012 |
| p Value | 0.0015 ** | 0.4639 | 0.7746 | 0.0048 * | 0.3535 |
were pooled into a single group) occurred at the frequency of 7.6%. Block 2 was composed of two

### 2.4.2. Haplotype Regression Analyses

Pair-wise D' values and inferred haplotype blocks were shown in Figure 3. For the six SNPs, two adjacent blocks were identified. Block 1 consisted of four SNPs, which formed four haplotypes in the studied population. The major haplotypes CACT, TATT and TGTT accounted for the frequency of 34.8%, 31.9% and 25.7%, respectively. However, the pooled haplotypes (which with frequency <5% were pooled into a single group) occurred at the frequency of 7.6%. Block 2 was composed of two SNPs, which formed three haplotypes.

### Table 3. Cont.

| SNP     | Genotype | MY     | FY     | FP     | PY     | PP      |
|---------|----------|--------|--------|--------|--------|---------|
| c.-1762 T>C | CC (71)  | 620.74 ± 97.86 ^A | 15.84 ± 4.20 ^A,a | −0.05 ± 0.044 | 19.10 ± 2.94 ^A | −0.009 ± 0.015 |
|         | CT (240) | 432.42 ± 75.59 ^A | 7.81 ± 3.41 ^B,b | −0.06 ± 0.035 | 14.93 ± 2.39 ^A | 0.006 ± 0.010 |
|         | TT (185) | 237.05 ± 77.76 ^B | 3.67 ± 3.47 ^Bb | −0.04 ± 0.036 | 9.35 ± 2.43 ^B | 0.012 ± 0.011 |
|         | p Value  | <0.0001 ** | 0.0013 ** | 0.6539 | <0.0001 ** | 0.2924 |
| c.-1621 T>G | GG (48)  | 500.63 ± 110.74 | 11.59 ± 4.68 | −0.04 ± 0.050 | 17.44 ± 1.19 ^A | 0.013 ± 0.017 |
|         | GT (186) | 240.61 ± 77.84 | 4.37 ± 3.48 | −0.03 ± 0.036 | 8.99 ± 1.69 ^B | 0.013 ± 0.011 |
|         | TT (262) | 312.57 ± 74.72 | 6.77 ± 3.38 | −0.03 ± 0.034 | 10.39 ± 0.92 ^B | 0.005 ± 0.010 |
|         | p Value  | 0.0279 | 0.1578 | 0.9389 | 0.0777 ** | 0.6157 |
| c.-1376 G>A | AA (14)  | 115.65 ± 167.77 ^A | −2.48 ± 6.82 ^A | −0.05 ± 0.075 | 3.64 ± 4.77 ^A | 0.007 ± 0.026 |
|         | AG (162) | 158.20 ± 79.24 ^A | 1.22 ± 3.52 ^A | −0.03 ± 0.036 | 4.86 ± 2.47 ^A | 0.003 ± 0.011 |
|         | GG (320) | 459.43 ± 73.27 ^B | 9.06 ± 3.33 ^B | −0.05 ± 0.033 | 14.71 ± 2.33 ^B | 0.004 ± 0.010 |
|         | p Value  | <0.0001 ** | 0.0009 ** | 0.6907 | <0.0001 ** | 0.9834 |
| c.-724 A>G | AA (185) | 547.23 ± 79.37 ^A | 10.08 ± 2.93 ^A | −0.07 ± 0.036 | 16.35 ± 2.49 ^A | −0.003 ± 0.011 |
|         | AG (262) | 282.30 ± 73.85 ^B | 5.26 ± 2.75 ^Bb | −0.03 ± 0.034 | 9.65 ± 2.34 ^B | 0.009 ± 0.010 |
|         | GG (49)  | 362.21 ± 106.47 ^B | −3.01 ± 3.37 ^B | −0.03 ± 0.048 | 0.50 ± 3.15 ^C | 0.001 ± 0.016 |
|         | p Value  | <0.0001 ** | 0.0016 ** | 0.2956 | <0.0001 ** | 0.3851 |

Bonferroni corrected significant level at p < 0.05 and p < 0.01 were 0.0083 and 0.0017, respectively; * p indicates the significant association after Bonferroni correction for multiple testing at the significance level of 0.05. ** p indicates the significant association after Bonferroni correction for multiple testing at the significance level of 0.01; ^A,b within the same column with different superscripts means p < 0.05, ^A,C means p < 0.01.

### Table 4. Genetic effects of these SNPs on the five traits in dairy cattle. Notes: * p < 0.05; ** p < 0.01.

| Locus   | Gene Effects | MY     | FY     | FP     | PY     | PP      |
|---------|--------------|--------|--------|--------|--------|---------|
| c.-2005 A>G | Additive (a) | 169.64 ** | 5.26 ** | −0.006 | 4.39 ** | −0.009 |
|         | Dominant (d) | 1.24 | −2.35 | −0.022 | −0.68 | −0.005 |
|         | Allele substitution (α) | 47.04 ** | 4.57 ** | −0.012 | 4.21 ** | −0.010 |
| c.-1762 T>C | Additive (a) | −176.72 ** | −2.60 | 0.013 | −4.56 ** | −0.003 |
|         | Dominant (d) | −57.91 | −2.07 | 0.001 | −1.86 | 0.001 |
|         | Allele substitution (α) | −204.52 ** | −3.59 | 0.014 | −5.45 ** | −0.005 |
| c.-1762 T>C | Additive (a) | 191.84 ** | 6.09 ** | −0.005 | 4.87 ** | −0.011 |
|         | Dominant (d) | 3.33 | −1.94 | −0.018 | 0.71 | 0.004 |
|         | Allele substitution (α) | 192.65 ** | 5.64 ** | 0.009 | 5.03 ** | −0.010 |
| c.-1621 T>G | Additive (a) | 94.03 | 2.41 | −0.007 | 3.83 ** | 0.004 |
|         | Dominant (d) | −165.99 * | −4.81 * | 0.009 | −4.93 ** | 0.004 |
|         | Allele substitution (α) | 24.31 | 0.39 | −0.003 | 1.46 ** | −0.006 |
| c.-1376 G>A | Additive (a) | −172.88 * | −5.79 * | 0.001 | −5.54 ** | 0.001 |
|         | Dominant (d) | −128.34 | −2.08 | 0.021 | −4.32 | −0.003 |
|         | Allele substitution (α) | −252.53 ** | −7.06 ** | 0.014 | −8.22 ** | −0.001 |
| c.-724 A>G | Additive (a) | 253.51 ** | 1.90 ** | −0.021 | 8.03 ** | −0.002 |
|         | Dominant (d) | −9.42 | 2.37 | 0.017 | 1.12 | 0.010 |
|         | Allele substitution (α) | 252.87 ** | 2.28 ** | −0.016 | 8.34 ** | 0.001 |

(a), (d), (α), means Additive, Dominant, Allele substitution effects, respectively; * the bold number means significant at p < 0.05; ** the bold number means significant at p < 0.01.
The single-marker analysis is usually thought to be noisy and less powerful due to lack of the simultaneous use of multi-point information [17,18]. Haplotype analysis, unlike single-marker analysis, is more likely to have significant effects on traits [19,20]. As shown in Table 5, the haplotypes in the two blocks both showed a significant association with MY, FY and PY. The results of haplotype association analyses were mostly consistent with those of the single-locus, which provided strong evidence for the associations between these SNPs and haplotypes with milk production traits.

2.5. Functional Prediction of the Allele-Dependent TFBS

As the most important mechanisms, the mutations in the regulatory region of a gene can affect transcription rate by changing the transcription factor binding sites (TFBSs) [21]. In view of the importance of these mutations, our study has highlighted the functions of these variants and their involvement in gene expression. Bioinformatic analyses using the TFSEARCH software revealed that the two regulatory SNPs, c.-1376 G>A and c.-724 A>G in PDE9A were predicted to change the binding site of the corresponding transcript factor acute myeloid leukemia 1 (AML-1a) and upstream stimulatory factor (USF), respectively (Figure 4). USF family proteins are well known as ubiquitously expressed transcription factors with inhibition effect [22]. Functional analysis in cultured mammalian cells indicated that USFs are involved in the gene networks of cell proliferation [23]. However, none of reports on the biological function of AML-1a have been found.
2.6. Expression Regulation of the Mutations in PDE9A Gene

To investigate potential regulation of the SNPs in 5’ regulatory region, the mRNA expression of PDE9A was measured in mammary glands based on the genotypes for both c.-1376 G>A and c.-724 A>G. The results showed that the mRNA levels with GG genotype of c.-1376 G>A and AA genotype of c.-724 A>G were higher than those of the other genotypes in mammary gland of Holstein cows in lactation (p < 0.01, p < 0.05; Figures 5 and 6), respectively. In this study, the higher expression levels of PDE9A are also consistent with the higher EBVs for milk production traits. Hence, we hypothesized that the effects of such SNPs on the milk production traits may be partly induced by their regulatory role on the expression of PDE9A.

Figure 4. Prediction the alteration of TFBSs in the mutations c.-1376 G>A and c.-724 A>G by using TFSEARCH. The letters in red indicate the position of mutations; the arrows mean the direction of TFBSs.

Figure 5. mRNA expression of c.-1376 G>A among the different genotypes in the mammary gland of lactating Holstein cows. Bars represent (mean value ± standard error) (n = 4). ** p < 0.01.

Figure 6. mRNA expression of c.-724 A>G among the different genotypes in the mammary gland of lactating Holstein cows. Bars represent mean ± SE (n = 5, 2, and 1, respectively). * p < 0.05.
In this study, we demonstrated that significant associations exist between \textit{PDE9A} variants and milk production traits in Chinese Holstein cows. To find more evidence for such associations, we compared our results with the known major gene and previously reported QTL and GWAS data. \textit{PDE9A} is 2.8–4.0 Mb away from the three significant SNPs for fat yield, fat percentage and protein percentage with \( p \) value of \( 1.58 \times 10^{-6} \)–\( 8.26 \times 10^{-7} \) reported by the same GWAS [24]. It is also located within the QTL region of 127.4–148 cM for fat yield detected in a German Holstein cattle population [25].

As a novel candidate for milk production traits, \textit{PDE9A} is merely studies in humans and mice until now [26,27]. \textit{PDE9A} is a cGMP-specific enzyme and can activate cGMP-dependent signaling pathways by influencing intracellular cGMP levels [28]. As an important signaling molecule within cells, cGMP regulates cell growth, adhesion and energy homeostasis in normal physiological conditions [29]. Meanwhile, cGMP may be involved in the regulation of lipogenesis in mouse mammary gland [30]. Interestingly, a study showed that cGMP also participated in milk formation in goat mammary gland (unpublished data). In addition, \textit{PDE9A} was reported to be a major regulator of basal cGMP levels in human breast cancer cells and inhibition of phosphodiesterases elevated intracellular cGMP [8] and involved in the negative feedback control of cyclic nucleotides pathways [31]. According to the phylogenetic tree, the bovine \textit{PDE9A} was found to be phylogenetically closest to \textit{Capra hircus} and \textit{Ovis aries PDE9A}. Meanwhile, it was closely associated with the mouse and rat \textit{PDE9A}. Considering the conservation of \textit{PDE9A} gene in mammalian species, we inferred that the \textit{PDE9A} gene might potentially influence the formation of milk traits or mammary physiological processing. In this study, the higher expression levels of \textit{PDE9A} are also consistent with the higher EBVs for milk production traits.

In association analyses, the EBVs of daughters were used as phenotypic observations. EBVs are usually used as dependent variables in association analyses concerning milk production traits in dairy cattle [5,32]. We have also compared the results denoted by EBVs and phenotypic values for association analyses in our study, which demonstrated that the findings based on the two different variables were basically overlapped. Consequently, we herein used the EBVs for association analysis. As the haplotype analysis including multiple marker information was thought more powerful than single marker analysis [33], we further performed the haplotype association analysis to detect the association of \textit{PDE9A} variants with milk production traits. It is obviously clear that the results of single-locus and haplotype association analysis were totally consistent. Accordingly, our findings could be used as genetic marker in genomic selection program of dairy cattle.

Compared to conventional breeding, the marker-assisted selection (MAS) can accelerate genetic gains dramatically [34]. Based on our findings herein, significant associations between polymorphisms in \textit{PDE9A} and dairy production traits were revealed in dairy cattle. Specifically, we can incorporate the information of the \textit{PDE9A} into selection program for practical use of these variations in breeding programs. Further biological analysis should be conducted in order to check the effects of such polymorphisms and validate its function on milk production traits before apply as markers for marker-assisted selection in the Chinese Holstein population. In addition, the promising variants of the \textit{PDE9A} gene can be applied to increase the proportion of the marker, which is positively associated with the milk production traits.

3. Experimental Section

3.1. Bioinformatic Analysis of \textit{PDE9A} Protein

\textit{PDE9A} protein homology between the bovine and other species were obtained from Genbank using the BLAST program [35]. Multiple alignments were conducted using Clustal X Version 2.0. Moreover, a phylogenetic tree of \textit{PDE9A} protein was further constructed using the Molecular Evolutionary Genetics Analysis (MEGA5.1) software [36]. Based on the neighbor-joining method,
the number noted at branches indicates the percentage of times that a node was supported in 1000 bootstrap pseudo replications.

3.2. Gene Expression Assays of PDE9A Gene

To further confirm the potential function of PDE9A, we conducted gene expression analyses aiming at different tissues and different genotypes, respectively. Eight lactating Chinese Holstein cows in their 2nd/3rd lactation were selected from the Beijing F dairy farm center. Samples of eight tissues, including mammary gland, heart, uterus, kidney, liver, gluteus, ovary and small intestine, were collected for each cow and then stored at liquid nitrogen.

Total RNA was isolated from tissues using Trizol reagent (Invitrogen, Carlsbad, CA, USA) according to the manufacturer’s protocols. Subsequently, RNA extracts were incubated with RNase-free DNase I for 30 min at 37 °C to remove DNA contamination. Reverse transcriptase reaction was conducted using PrimerScriptHT RT reagent Kit (TaKaRa Biotechnology Co., Ltd., Dalian, China). Real-time PCR using SYBR green fluorescence (Roche, Penzberg, Germany) was performed with a volume of 15 µL containing 7.5 µL SYBR Green Mixture, 2 µL template of cDNA, 0.375 µL of each primer, 4.75 µL distilled water. The PCR conditions: denaturation at 95 °C for 10 s; amplification 55 cycles at 95 °C for 10 s, 59 °C for 15 s, followed by 72 °C for 25 s. Quantitative real-time PCR primers of PDE9A and GAPDH were shown in Table S1. All measurements were performed in triplicate and the relative gene expression was normalized by the GAPDH with 2^{-ΔΔCt} method, as described previously [16].

To further detect the effects of variants of PDE9A, the mRNA expression of mammary glands with different genotypes at functionally important mutations were also analyzed. On the basis of association analysis results and genotypes of the two SNPs (c.-1376 G>A, c.-724 A>G) in the 5' regulatory region, the PDE9A mRNA expression of mammary glands in eight daughters among different genotypes were also detected. The results of mRNA expression were performed by GLM procedure of the SAS 9.1.3 software (SAS Institute, Cary, NC, USA).

3.3. Resource Population and DNA Extraction

The resource population was chosen from 14 dairy farms in the Sanyuanlvhe Dairy Farming Center (Beijing, China), where routine standard performance test (Dairy Herd Improvement system, DHI, Beijing, China) have been carried out since 1999. A daughter design was employed in this study. A total of 506 Chinese Holstein cows, which were daughters from 8 sire families, were selected to construct a single population in this study. Estimated breeding values (EBVs) of recorded cattle for each milk production trait (i.e., milk yield, fat yield, protein yield, fat percentage, and protein percentage) were provided by the Dairy Association of China (DAC) using the genetic parameters estimated based on the complete DHI data of Chinese dairy cattle population.

Genomic DNA (gDNA) was isolated from whole blood samples of the daughters by Blood DNA Kit according to the manufacturer’s instructions (Tiangen Biotech Co., Beijing, China) and frozen semen of the 8 bulls using a standard phenol-chloroform method and stored at −20 °C. The quality and quantity of extracted genomic DNA were measured with NanoDrop™ Spectrophotometer (ND-2000c) (Thermo Scientific, Chelmsford, MA, USA).

3.4. SNP Identification and Genotyping

Genomic DNA samples of 8 Holstein bulls were selected to construct a pool DNA with identical DNA concentration (50 ng/mL). Using software Primer3.0, according to the genomic sequence of PDE9A (AC000158.1), a total of 31 pairs of primers were designed to amplify their entire coding regions, partial introns as well as regulatory regions of 5' and 3', respectively (Table S2). PCR amplification was performed in a volume of 50 µL, containing 2 µL of pooled DNA, 20 pmol of each forward and reverse primer, 25 µL of 2× MightyAmp buffer, and 1.25 U of MightyAmp DNA polymerase (Takara Biotechnology Co., Ltd., Dalian, China). The Cycling parameters were as follows:
denaturation 95 °C for 10 min, followed by 36 cycles of 95 °C for 30 s, 60 °C for 30 s, 30 s at 72 °C, and with a final extension completed at 72 °C for 7 min. Then, PCR products of each fragment were sequenced by an ABI3730XL sequencer (Applied Biosystems, Foster City, CA, USA).

SNApshot assay was applied for genotyping of the 6 SNPs in the 5′-regulatory region of PDE9A of each cow with Peak ABI PRISM® SNaPshot™ Multiplex Kit (Applied Biosystems, Foster City, CA, USA). In follow-up analyses, scanner software v1.0 (Applied Biosystems) was used to identify the genotype of each individual for all the SNPs.

3.5. Haplotype Analysis

To further estimate the LD extent for each pair of variants genotyped in the PDE9A gene, the software Haploview 4.2 (Broad Institute of MIT and Harvard, Cambridge, MA, USA) was performed [14]. Accordingly, haplotype blocks were also explored with the subject genotype data via Haploview based on the criterion of D'. In this analysis, the haplotype with frequency >5% was considered as a single haplotype, and those with frequency <5% were pooled into a distinguishable group. Association between each haplotype and the five milk production traits was performed in subsequent analyses.

3.6. Association Analysis

Pedigree information of the resource population was traced back for three generations to create the numerator relationship matrix. Kinship matrix (A-matrix) was constructed using MATLAB version 7.11 (R2010b). Based on the estimated variance-covariance matrices, the association of SNPs and haplotypes of PDE9A with the 5 milk production traits were evaluated using the mixed procedure in SAS 9.1.3 (SAS Institute, Cary, NC, USA) with the following mixed animal model [31]. A linear mixed model was fitted as follows:

\[ y = 1\mu + bx + Za + e \]  

where \( y \) is the vector of phenotypes (EBV) of all the cows, \( \mu \) is the general mean. \( x \) is the fixed effect vector of the SNP genotype or haplotype combination, \( b \) is the regression coefficient, \( Z \) is the incidence matrix of \( a \), \( a \) is the vector of residual additive effects with distribution of \( a \sim N(0, A\delta^2_a) \) (\( A \) is the polygenic relationship matrix among all the individuals, \( \delta^2_a \) is the additive genetic variance), and finally \( e \) denotes the residual errors with distribution of \( e \sim N(0, W\delta^2_e) \) (\( W \) is a diagonal matrix with the diagonal elements equal to \( 1/REL_{ij} \) and \( \delta^2_e \) is the residual error variance).

The additive (\( a \)), dominance (\( d \)) and allele substitution (\( \alpha \)) effects were estimated using the equation: \( a = (AA - BB)/2 \), \( d = AB - (AA + BB)/2 \) and \( \alpha = a + d \) (\( q - p \)), where AA or BB indicates the two homozygous genotypes, AB indicates heterozygous genotype, \( p \) and \( q \) are the allele frequencies of corresponding locus. In addition, bonferroni correction was performed for multiple t-testing in subsequent analyses.

3.7. Bioinformatics Analysis of Transcription Factor Binding Sites (TFBSs)

The online software TFSEARCH [37] was used to predict the putative differential TFBSs according to the mutations with standard settings for the highest similarity of the 6 regulatory SNPs in PDE9A gene.

4. Conclusions

In summary, we validated the significant associations of PDE9A gene identified in previous GWA study with milk production traits and showed that mRNA expression levels of the PDE9A gene were highest in mammary gland tissue of Chinese Holstein. Additionally, two SNPs in 5′ regulatory region were identified to be involved in the regulation of gene expression. These findings strongly suggest that the associated variants could be used as molecular markers in advanced marker-assisted selection to facilitate the breeding of desired production traits in the Chinese Holstein population.
Supplementary Materials: Supplementary materials can be found at http://www.mdpi.com/1422-0067/16/11/25976/s1.

Acknowledgments: This work was supported by the High Technology Research and Development Program of China (2013AA102504), Beijing Innovation Team of Technology System in the National Dairy Industry, National Natural Science Foundation of China (31072016), National Key Technologies R & D Program (2011BAD28B02, 2012BAD12B01, 10ZCKFNC02700), 948 Program (2011-G2A), Beijing Research and Technology Program (D12110003312001) and Program for Changjiang Scholar and Innovation Research Team in University (IRT1191).

Author Contributions: Conceived and designed the experiments: Dong-Xiao Sun, Sheng-Li Zhang and Qin Zhang; Performed the experiments: Xiao-Jun Bi analyzed the data: Shao-Hua Yang and Xiao-Jun Bi; Contributed to sample collection: Yan Xie and Cong Li wrote the paper: Shao-Hua Yang, Xiao-Jun Bi, and D.-X.S.

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

References
1. Madeja, Z.; Adamowicz, T.; Chmurzynska, A.; Jankowski, T.; Melonek, J.; Switonski, M.; Strabel, T. Short communication: Effect of leptin gene polymorphisms on breeding value for milk production traits. J. Dairy Sci. 2004, 87, 3925–3927. [CrossRef]
2. Sahana, G.; Guldbrandtsen, B.; Thomsen, B.; Holm, L.E.; Panitz, F.; Brondum, R.F.; Bendixen, C.; Lund, M.S. Genome-wide association study using high-density single nucleotide polymorphism arrays and whole-genome sequences for clinical mastitis traits in dairy cattle. J. Dairy Sci. 2014, 97, 7258–7275. [CrossRef] [PubMed]
3. Scott, L.J.; Mohlke, K.L.; Bonnyycastle, L.L.; Willer, C.J.; Li, Y.; Duren, W.L.; Erdos, M.R.; Stringham, H.M.; Chines, P.; Jackson, A.U.; et al. A genome wide association study of type 2 diabetes in Finns detects multiple susceptibility variants. Science 2007, 316, 1341–1345. [CrossRef] [PubMed]
4. Jiang, L.; Liu, J.F.; Sun, D.X.; Ma, P.P.; Ding, X.D.; Yu, Y.; Zhang, Q. Genome wide association studies for milk production traits in Chinese Holstein population. PLoS ONE 2010, 5, e13661. [CrossRef] [PubMed]
5. Kolb, D.; Wang, Z.; Grant, J.R.; Murdoch, B.; Prasad, A.; Xiu, Z.; Marques, E.; Stothard, P.; Moore, S.S. A whole genome scan to map QTL for milk production traits and somatic cell score in Canadian Holstein bulls. J. Anim. Breed. Genet. 2009, 126, 216–227. [CrossRef] [PubMed]
6. Boichard, D.; Grohs, C.; Bourgeois, F.; Cerqueira, F.; Faugeras, R. Detection of genes influencing economic traits in three French dairy cattle breeds. Genet. Sel. Evol. 2003, 35, 77–101. [CrossRef] [PubMed]
7. Kroker, K.S.; Mathis, C.; Marti, A.; Cassel, J.C.; Rosenbrock, H.; Dorner-Ciossek, C. PDE9A inhibition rescues amyloid β-induced deficits in synaptic plasticity and cognition. Neurobiol. Aging 2014, 35, 2072–2078. [CrossRef] [PubMed]
8. Francis, S.; Blount, M.; Corbin, J. Mammalian cyclic nucleotide phosphodiesterases: Molecular mechanisms and physiological functions. Physiol. Rev. 2011, 91, 651–690. [CrossRef] [PubMed]
9. Conti, M.; Jin, S.L. The molecular biology of cyclic nucleotide phosphodiesterases. Prog. Nucleic Acid Res. Mol. Biol. 2000, 63, 1–38.
10. Houslay, M.D. PDE4 cAMP-specific phosphodiesterases. Prog. Nucleic Acid Res. 2001, 69, 249–315.
11. Aronsen, L.; Orvell, E.; Lysaa, R.; Ravna, A.W.; Sager, G. Modulation of high affinity ATP-dependent cyclic nucleotide transporters by specific and non-specific cyclic nucleotide phosphodiesterase inhibitors. Eur. J. Pharmacol. 2014, 745, 249–253. [CrossRef] [PubMed]
12. Guipponi, M.; Scott, H.S.; Kudoh, J.; Kawasaki, K.; Shibuya, K.; Shintani, A.; Asakawa, S.; Chen, H.; Lalioti, M.D.; Rossier, C.; et al. Identification and characterisation of a novel cyclic nucleotide phosphodiesterase gene (PDE9A) that maps to 21q22.3: Alternative splicing of mRNA transcripts, genomic structure and sequence. Hum. Genet. 1998, 103, 386–392. [CrossRef] [PubMed]
13. Kleiman, R.J.; Chapin, D.S.; Christoffersen, C.; Freeman, J.; Fonseca, K.R.; Geoghegan, K.F.; Grimwood, S.; Guanowski, V.; Hajos, M.; Harms, J.F.; et al. Phosphodiesterase 9A regulates central cGMP and modulates responses to cholinergic and monoaminergic perturbation in vivo. J. Pharmacol. Exp. Ther. 2012, 341, 396–409. [CrossRef] [PubMed]
14. Cases, S.; Smith, S.J.; Zheng, Y.W.; Myers, H.M.; Lear, S.R.; Sande, E.; Novak, S.; Collins, C.; Welch, C.B.; Lusis, A.J.; et al. Identification of a gene encoding an acyl CoA: Diacylglycerol acyltransferase, a key enzyme in triacylglycerol synthesis. Cell Biol. 1998, 95, 131018–131023.
15. Bionaz, M.; Loor, J.J. Gene networks driving bovine milk fat synthesis during the lactation cycle. *BMC Genom.* 2008, 9, 366. [CrossRef] [PubMed]

16. Lengi, A.J.; Corl, B.A. Identification and characterization of a novel bovine stearoyl-CoA desaturase isozyme with homology to human SCD5. *Lipids* 2007, 42, 499–508. [CrossRef] [PubMed]

17. Daly, M.J.; Rioux, J.D.; Schaffner, S.F.; Hudson, T.J.; Lander, E.S. High-resolution haplotype structure in the human genome. *Nat. Genet.* 2001, 29, 229–232. [CrossRef] [PubMed]

18. Huang, Q.; Fu, Y.X.; Boerwinkle, E. Comparison of strategies for selecting single nucleotide polymorphisms for case/control association studies. *Hum. Genet.* 2003, 113, 253–257. [CrossRef] [PubMed]

19. Capparelli, R.; Parlato, M.; Amoroso, M.G.; Roperto, S.; Marabelli, R.; Roperto, F.; Iannelli, D. Mannose-binding lectin haplotypes influence Brucella abortus infection in the water buffalo (*Bubalus bubalis*). *Immunogenetics* 2008, 60, 157–165. [CrossRef] [PubMed]

20. Abdel-Shafy, H.; Bortfeldt, R.H.; Tetens, J.; Brockmann, G.A. Single nucleotide polymorphism and haplotype effects associated with somatic cell score in German Holstein cattle. *Genet. Sel. Evol.* 2013, 46, 35. [CrossRef] [PubMed]

21. Wang, X.; Tomso, D.J.; Liu, X.; Bell, D.A. Single nucleotide polymorphism in transcriptional regulatory regions and expression of environmentally responsive genes. *Toxicol. Appl. Pharmacol.* 2005, 207, S84–S90. [CrossRef] [PubMed]

22. Sirito, M.; Lin, Q.; Maity, T.; Sawadogo, M. Ubiquitous expression of the 43- and 44-kDa forms of transcription factor USF in mammalian cells. *Nucleic Acids Res.* 1994, 22, 427–433. [CrossRef] [PubMed]

23. Patricia, G.S.; Antonio, F.P.; Mario, V. Differential configurations involving binding of USF transcription factors and Twist1 regulate Alx3 promoter activity in mesenchymal and pancreatic cells. *Biochem. J.* 2013, 450, 199–208.

24. Cole, J.B.; Wiggans, G.R.; Ma, L.; Sonstegard, T.S.; Lawlor, T.J. Genome-wide association analysis of thirty one production, health, reproduction and body conformation traits in contemporary U.S. Holstein cows. *BMC Genom.* 2011, 12, 408. [CrossRef] [PubMed]

25. Harder, B.; Bennewitz, J.; Reinsch, N.; Thaller, G.; Thomsen, H.; Kuhn, C.; Schwerin, M.; Erhardt, G.; Forster, M.; Reinhardt, F.; et al. Mapping of quantitative trait loci for lactation persistency traits in German Holstein dairy cattle. *J. Anim. Breed. Genet.* 2006, 123, 89–96. [CrossRef] [PubMed]

26. Wunder, F.; Tersteegen, A.; Rebmann, A.; Erb, C.; Fahrig, T.; Hendrix, M. Characterization of the first potent and selective PDE9 inhibitor using a cGMP reporter cell line. *Mol. Pharmacol.* 2005, 68, 1775–1781. [CrossRef] [PubMed]

27. Reneerkens, O.A.; Rutten, K.; Steinbusch, H.W.; Blokland, A.; Prickaerts, J. Selective phosphodiesterase inhibitors: A promising target for cognition enhancement. *Psychopharmacology* 2009, 202, 419–443. [CrossRef] [PubMed]

28. Andreeva, S.G.; Dikkes, P.; Epstein, P.M.; Rosenberg, P.A. Expression of cGMP-specific phosphodiesterase 9A mRNA in the rat brain. *J. Neurosci.* 2001, 21, 9068–9076.

29. Alexandra, M.F.; Gary, A.P.; Heather, N.T. The role of cyclic nucleotide signaling pathways in cancer: Targets for prevention and treatment. *Cancers* 2014, 6, 436–458.

30. Cameron, C.M.; Rillema, J.A. Effects of cyclic nucleotides on lipid biosynthesis in mouse mammary gland explants. *Proc. Soc. Exp. Biol. Med.* 1983, 173, 306–311. [CrossRef] [PubMed]

31. Gregor, P.D.; Sawadogo, M.; Roeder, R.G. The adenovirus major late transcription factor USF is a member of the helix-loop-helix group of regulatory proteins and binds to DNA as a dimer. *Genes Dev.* 1990, 4, 1730–1740. [CrossRef] [PubMed]

32. Daetwyler, H.D.; Schenkel, F.S.; Sargolzaei, M.; Robinson, J.A. A genome scanto detect quantitative trait loci for economically important traits in Holstein cattle using two methods and a dense single nucleotide polymorphism map. *J. Dairy Sci.* 2008, 91, 3225–3236. [CrossRef] [PubMed]

33. Martin, E.R.; Lai, E.H.; Gilbert, J.R.; Rogala, A.R.; Afshari, A.J.; Riley, J.; Finch, K.L.; Stevens, J.F.; Livak, K.J.; Slotterbeck, B.D.; et al. SNPing away at complex diseases: Analysis of single-nucleotide polymorphisms around APOE in Alzheimer disease. *Am. J. Hum. Genet.* 2000, 67, 383–394. [CrossRef] [PubMed]

34. Nguyen, M.T.; Andrew, C.B.; Peter, B.M.; Yutao, L.; Russell, E.L. Single nucleotide polymorphisms in the actin and crustacean hyperglycemic hormone genes and their correlation with individual growth performance in giant freshwater prawn *Macrobrachium rosenbergii*. *Aquaculture* 2010, 301, 7–15.

35. BLAST Program. Available online: http://blast.ncbi.nlm.nih.gov/Blast.cgi (accessed on 29 July 2014).
36. Kumar, S.; Nei, M.; Dudley, J.; Tamura, K. MEGA: A biologist-centric software for evolutionary analysis of DNA and protein sequences. *Brief Bioinform.* **2008**, *9*, 299–306. [CrossRef] [PubMed]

37. TFSEARCH. Available online: http://www.cbrc.jp/research/db/TFSEARCH.html (accessed on 10 September 2014).

© 2015 by the authors; licensee MDPI, Basel, Switzerland. This article is an open access article distributed under the terms and conditions of the Creative Commons by Attribution (CC-BY) license (http://creativecommons.org/licenses/by/4.0/).