

Genome-wide association study for the free amino acid and nucleotide components of breast meat in an F₂ crossbred chicken population

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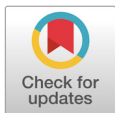
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Received: Sep 15, 2022
 Revised: Oct 26, 2022
 Accepted: Oct 31, 2022

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Abstract

Flavor is an important sensory trait of chicken meat. The free amino acid (FAA) and nucleotide (NT) components of meat are major factors affecting meat flavor during the cooking process. As a genetic approach to improve meat flavor, we performed a genome-wide association study (GWAS) to identify the potential candidate genes related to the FAA and NT components of chicken breast meat. Measurements of FAA and NT components were recorded at the age of 10 weeks from 764 and 767 birds, respectively, using a White leghorn and Yeonsan ogye crossbred F₂ chicken population. For genotyping, we used 60K Illumina single-nucleotide polymorphism (SNP) chips. We found a total of nine significant SNPs for five FAA traits (arginine, glycine, lysine, threonine content, and the essential FAAs) and one NT trait (inosine content), and six significant genomic regions were identified, including three regions shared among the essential FAAs, arginine, and inosine content traits. A list of potential candidate genes in significant genomic regions was detected, including the *KCNRG*, *KCNIP4*, *HOXA3*, *THSD7B*, and *MMUT* genes. The essential FAAs had significant gene regions the same as arginine. The genes related to arginine content were involved in nitric oxide metabolism, while the inosine content was possibly affected by insulin activity. Moreover, the threonine content could be related to methylmalonyl-CoA mutase. The genes and SNPs identified in this study might be useful markers in chicken selection and breeding for chicken meat flavor.

Keywords: Chicken, Breast meat, Free amino acids, Nucleotides, Genome-wide association study

INTRODUCTION

Yeonsan ogye is a chicken breed native to Korea. The nutritional quality, taste, overall acceptance, and

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Competing interests

No potential conflict of interest relevant to this article was reported.

Funding sources

This work was carried out with the support of "Cooperative Research Program for Agriculture Science & Technology Development (Project No. PJ0157852022)" Rural Development Administration, Korea and also supported by Chungnam National University.

Acknowledgements

Not applicable.

Availability of data and material

Upon reasonable request, the datasets of this study can be available from the corresponding author.

Authors' contributions

Conceptualization: Cha J, Jin D, Lee JH.
 Data curation: Kim M, Cho E, Cha J, Jin D, Cho S.
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 Writing - original draft: Kim M.
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Ethics approval and consent to participate

Experimental use of animal was approved by Animal Ethics Committee of Chungnam National University (202103A-CNU-061).

texture of the meat of the Korean native chicken breed, including the content of taste-related compounds, are superior to those of commercial broilers [1,2]; thus, consumers have a higher preference for it. Therefore, it would be a good strategy to crossbreed native chickens to produce broilers with tastier meat by taking advantage of the unique meat quality characteristics of native chickens. A study of the characteristics of the Yeonsan ogye chicken meat reported higher crude fat content and shearing power in Yeonsan ogye compared to other Korean native chickens, and the color of the meat was expressed as black [3]. In addition, it has been reported that the breast meat extract of Korean native chickens can alleviate inflammation [4].

To improve the palatability of chicken products, many studies have been conducted to determine the factors that influence meat quality. Studies have identified several parameters that influence meat quality, including tenderness, water-holding capacity, and meat color, in relation to the physical composition, while pH, dry matter, moisture, and intramuscular fat content are related to the chemical composition of the meat [5–8]. Taste-active precursors in meat, such as free amino acid (FAA) and nucleotide (NT), undergo chemical reactions during cooking, which determines many aspects of flavor, one of the main sensory features of meat [9–11].

As reported in [9], glycine, alanine, serine, threonine, and proline in the form of FAAs provide the sweet taste to the meat after cooking. Regarding NTs, inosine 5'-monophosphate, alone or in combination with other flavor compounds, provides an umami taste, greatly affecting the flavor of meat [12].

Many studies have been conducted to identify the genes related to the physicochemical properties of chickens. In [13], a study of the association of the fatty acid composition in chicken was conducted, and in [14], a gene related to the lipid profile of chicken muscle was reported. In [15,16], association studies were conducted on body composition. However, there have been no attempts to identify the genes related to FAA and NT content, which determine the flavor of chicken, in the whole genome. Furthermore, the traits present in the chicken Quantitative Trait Loci Database do not include the FAA and NT contents of breast meat [17]. Therefore, the objective of this study was to identify the potential candidate genes related to different FAA and NT contents in chicken breast through a whole-genome association analysis.

MATERIALS AND METHODS

Experimental animals

For the construction of the experimental chicken population, a reciprocal crossbred F_2 generation was produced using the Yeonsan ogye breed and the White Leghorn breed owned by the National Institute of Animal Science (NIAS, Korea). A total of 810 birds from the F_2 generation were made up of two lines according to the sex combination of F_0 birds and were raised until 10 weeks of age with the mixing of males and females. All F_2 birds were raised on a farm at the NIAS Center for Livestock Genetic Resources under the same environmental and nutritional conditions. Overall experimental outline and the construction of the F_2 generation is illustrated in Fig. 1.

Phenotype measurement and quality control

At 10 weeks of age, all birds were slaughtered on the same day at a commercial abattoir. Viscera were removed from the chicken carcass immediately after slaughter, rapidly frozen at -35°C , and then stored at -20°C at NIAS (Wanju, Korea) on the same day. Frozen samples were stored for 2–6 weeks, and then transported to Chungnam National University (Deajeon, Korea) and stored at -80°C until deboning. Next, samples were thawed at 4°C for 20 h, the breast meat was separated from the carcass, and then the breast meat was divided into samples for the FAA and

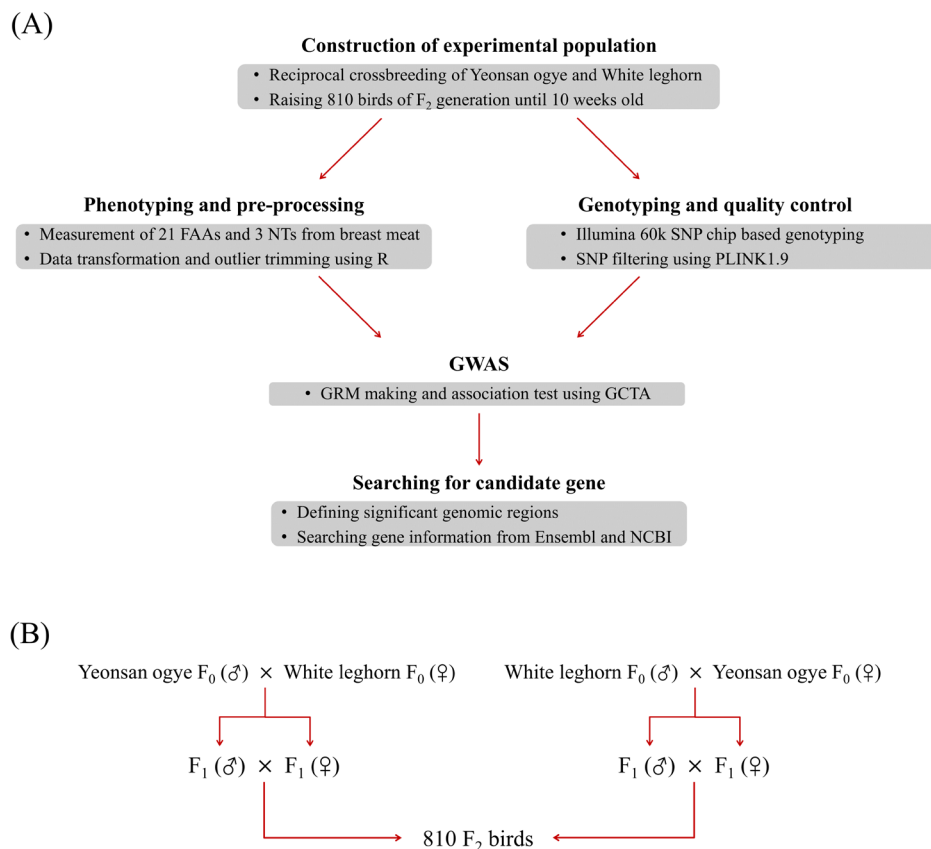


Fig. 1. Outline of the experimental workflow and the construction of the F₂ generation chicken population. (A) Diagram of overall experimental contents. (B) The construction of the reciprocal crossbred F₂ generation chickens using Yeosan ogye and White Leghorn. The male bird of Yeosan ogye and the female bird of White Leghorn produced F₁ generation birds. And F₁ birds mated between F₁ siblings so that a part of F₂ birds was generated. Another part of F₂ birds was made up of the F₂ progeny of a different sex combination of Yeosan ogye and White Leghorn chickens as the F₀ generation. FAA, free amino acid content; NT, nucleotide content; SNP, single nucleotide polymorphism; GWAS, genome-wide association study; GRM, genetic relationship matrix; GCTA, genome-wide complex trait analysis.

NT component analyses. The FAA analysis was performed using a liquid chromatography system (UltiMate 3000RS, Thermo Fisher Scientific, Waltham, MA, USA) to quantify the content of 21 FAAs from a sample (Table 1). From the quantified FAA results, the essential FAAs and non-essential FAAs contents of chickens were added and each phenotype was used in an association analysis. For the NT content analysis, high-performance liquid chromatography (1200 Series, Agilent, Santa Clara, CA, USA) was applied. The resulting phenotypes were used in the association test as traits for the content of each substance and are summarized in Table 1.

To normalize the phenotypic data, a transformation was carried out to accurately estimate the effect of single-nucleotide polymorphisms (SNPs) in the association test. For each phenotype in the dataset, a Shapiro–Wilk normality test was performed using R software with the “shapiro.test()” function, and a transformation step was applied to the phenotype with a p -value of < 0.05 . To find the optimal transformation method, the dataset with the highest p -value was adopted after performing normality tests on the raw data and data transformed by log, square root, square, and minimum/maximum (min/max) methods (Table 1). Then, to remove outliers, values of three standard deviations or more from the mean of each data point were treated as missing data, and all subjects whose analysis results were identified as missing data were excluded from the analysis. The

Table 1. Descriptive statistics of the measured phenotypes in the F₂ chicken population

Trait	Type	unit	Mean	SD	CV (%)	Min	Max	Trans
Alanine	Nonessential	μmol/g	1.5772	0.7439	0.4716	0.032	5.139	Sqrt
Arginine	Essential	μmol/g	9.0896	5.9595	0.6556	0.059	42.62	Sqrt
Asparagine	Nonessential	μmol/g	0.2946	0.158	0.5363	0	0.834	Sqrt
Aspartic acid	Nonessential	μmol/g	0.4188	0.2583	0.6169	0.012	1.93	Sqrt
Cysteine	Nonessential	μmol/g	3.4547	1.9151	0.5544	0.09	12.61	Sqrt
GABA	Nonessential	μmol/g	0.0564	0.0632	1.1207	0	0.68	Sqrt
Glutamic acid	Nonessential	μmol/g	1.2466	0.661	0.5302	0.031	9.389	Sqrt
Glutamine	Nonessential	μmol/g	1.1161	0.6518	0.584	0.02	5.257	Sqrt
Glycine	Nonessential	μmol/g	0.8388	0.5757	0.6863	0.01	4.91	Sqrt
Histidine	Essential	μmol/g	0.34	0.1819	0.5349	0.01	1.07	Sqrt
Isoleucine	Essential	μmol/g	0.3973	0.1789	0.4503	0.008	1.12	Min-Max
Leucine	Essential	μmol/g	0.8782	0.4067	0.4631	0.019	2.38	No trans
Lysine	Essential	μmol/g	0.0299	0.1128	3.7769	0	1.44	Sqrt
Methionine	Essential	μmol/g	0.6394	0.5242	0.8198	0.013	4.169	Log
Phenylalanine	Essential	μmol/g	0.4195	0.1998	0.4762	0.006	1.58	Min-Max
Proline	Nonessential	μmol/g	0.3564	0.2792	0.7834	0.004	6.29	Sqrt
Serine	Nonessential	μmol/g	1.342	0.5624	0.419	0.039	3.26	Min-Max
Threonine	Essential	μmol/g	0.8727	0.4369	0.5006	0.014	3.007	Sqrt
Tryptophan	Essential	μmol/g	0.0343	0.0376	1.0964	0	0.89	Sqrt
Tyrosine	Nonessential	μmol/g	0.6248	0.5362	0.8582	0.019	2.97	Log
Valine	Essential	μmol/g	0.3802	0.2194	0.5771	0.002	1.707	Sqrt
Essential FAAs		μmol/g	15.2358	7.4925	0.4918	0.32	52.59	Sqrt
Nonessential FAAs		μmol/g	9.108	3.7094	0.4073	0.74	25.98	Sqrt
AMP	NT	mg/100g	10.2654	4.4131	0.4299	0.03	30.61	Sqrt
IMP	NT	mg/100g	297.7067	86.4317	0.2903	24.4	572.61	No trans
Inosine	NT	mg/100g	49.8131	20.7431	0.4164	4.6749	146.47	Sqrt
Hypoxanthine	NT	mg/100g	15.7305	5.7408	0.3649	4.74	42.52	Log

CV, coefficient of variation; Min, minimum; Max, maximum; Trans, transformation; Sqrt, square root scaling; GABA, γ-aminobutyric acid; Min-Max, minimum-maximum scaling; Log, Log scaling; FAA, free amino acid content; AMP, adenosine monophosphate; NT, nucleotide content; IMP, inosine 5'-monophosphate.

FAA and NT component datasets were filtered in the quality control process, leaving 764 and 767 birds for the later analysis of FAA and NT components, respectively.

Genotype data generation and quality control

For genomic DNA (gDNA) extraction, blood collection was performed from the brachial veins of birds at 8 weeks of age. gDNA was extracted from the collected blood using the Wizard Genomic DNA Purification Kit (Promega, Madison, WI, USA), and the quality and concentration of the DNA were checked using a NanoDrop 1000 spectrophotometer (Thermo Fisher Scientific), with samples stored at -20°C. A total of 57,636 SNP genotypes were produced from the extracted gDNA using the Illumina 60K Chicken SNP BeadChip. The SNPs were removed if any of the following conditions were not met: call rate > 0.9, minor allele frequency > 0.01, Hardy-Weinberg equilibrium test *p*-value of > 0.01 × 10⁻⁶. After filtering, 29,175 SNPs were available for the subsequent analysis. PLINK 1.9 software was used for genotype quality control procedure [18].

The genome-wide association study (GWAS) and identification of the candidate genes

A genetic relationship matrix (GRM) was calculated using the genome-wide complex trait analysis (GCTA) tool[19]. An association analysis was performed using GCTA's mixed linear model leave-one-chromosome-out statistical method, and the following model was used:

$$y = a + bx + \bar{g} + e$$

where, y is a vector of the phenotype value, including FAAs and NTs; a is the mean value of phenotypic record; b is the additive effect of the SNP to be tested for association; x is the genotype of the SNP to be tested; \bar{g} is the effect of all SNPs captured by the GRM, except those on the chromosome where SNP testing is located; and e is the residual effect vector. To accurately estimate the marker effect, sex, experimental batch, half-sibling family, and body weight were used as covariates. After the association analysis, the significance of the SNP effect was evaluated by a Bonferroni corrected p -value of 5%: $p < 0.05/\text{number of SNPs use for the association test}$. The GWAS result was visualized with a Manhattan plot using R software.

Based on the chicken reference genome (GRCg6a), we assumed that the genomic region around 1 Mbp was the significant region centered on the position of the most significant SNP for each chromosome, and known genes included in that region were reported as candidate genes. Location-based gene information was obtained by searching the Ensembl (<https://asia.ensembl.org>) and NCBI (<https://www.ncbi.nlm.nih.gov>) databases.

RESULTS

Basic statistics of phenotypic values

The descriptive statistics of the phenotype measurements of FAAs and NTs from the F₂ chicken population are presented in Table 1. For these phenotypic values, basic statistical analyses were performed after transformation and outlier trimming. Histograms and quantile–quantile plots are shown in Fig. 2. Lysine, for which most values were measured as zero, was excluded from the test because its distribution was judged to be inappropriate for the association test. It was determined that the cause of this inappropriate data was an error in phenotype measurement.

GWAS results and related genomic regions

Table 2 shows the significant SNPs for each trait as a result of the association test, with results presented as summary statistics of the GWAS. The SNP markers were located at chicken (*Gallus gallus*) chromosome 1 (GGA1), GGA2, GGA3, GGA4, and GGA7. Of the three SNPs associated with the essential FAAs, Gga_rs15943371 and Gga_rs14162297 were located at bp 33,096,495 and 33,432,575 on GGA2, respectively, and Gga_rs16609168 was located at bp 31,241,647 on GGA7. In arginine, Gga_rs14162297 did not reach the significance level, but it was the second most significant SNP among the markers located on GGA2, and Gga_rs16609168 was also the SNP with the highest significance in GGA7. Another SNP related to arginine was Gga_rs13971906, which was located at bp 170,582,107 on GGA1, and was significant for inosine. As other SNPs associated with inosine, GGaluga265714, GGaluga265712, and GGaluga265969 were located on GGA4 at bp 74,314,272, 74,285,145, and 75,827,200, respectively. The SNP associated with glycine was GGaluga028030 and was located at bp 81,831,988 on GGA1. The SNP associated with threonine was Gga_rs14412837 and was located at bp 108,537,571 on GGA3.

The Manhattan plot in Fig. 3 presents only the traits with significant SNPs. The most significant

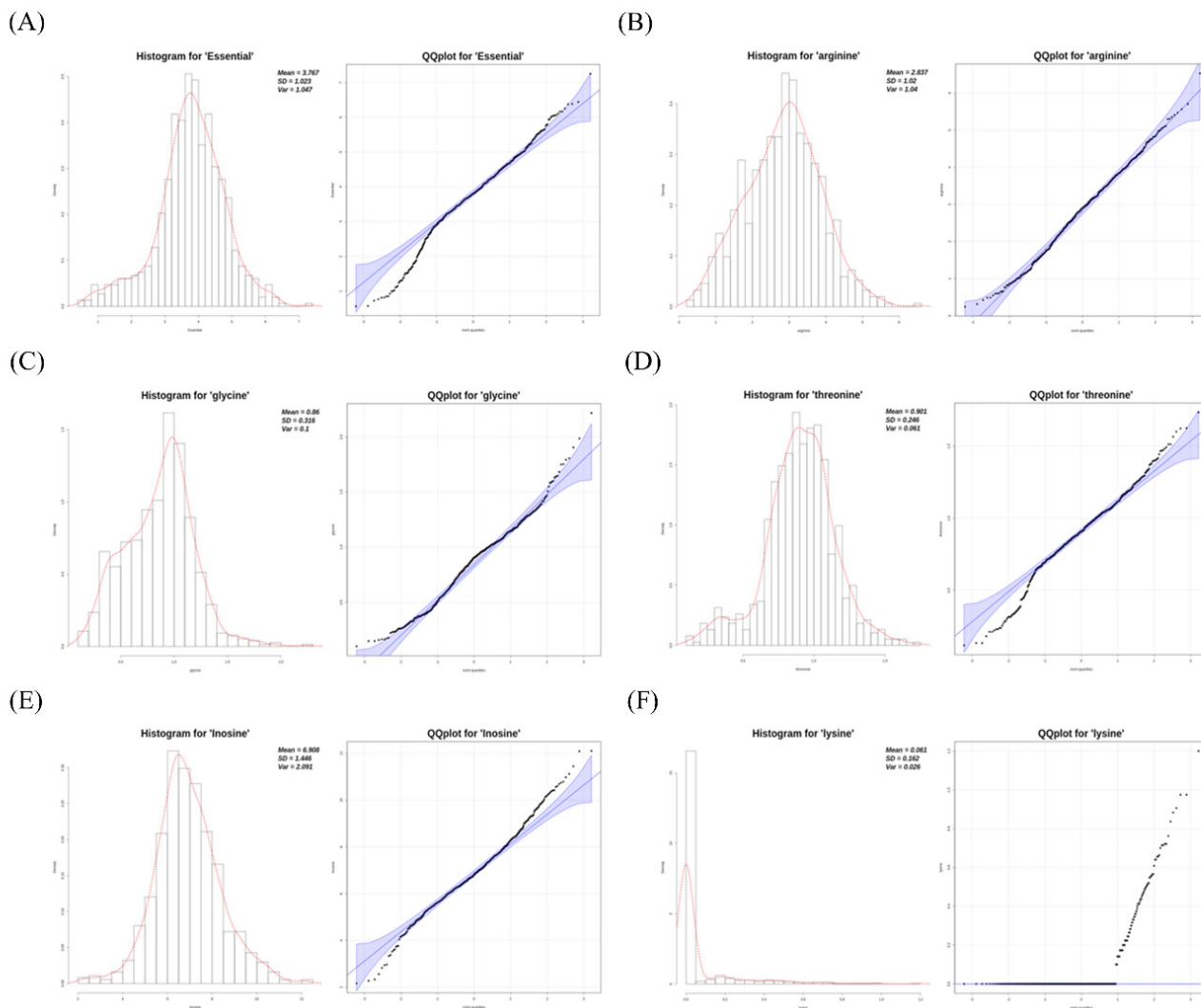


Fig. 2. Phenotype distribution of free amino acids and nucleotide contents used for the association test. Histogram and quantile–quantile plot contents, respectively: (A) essential free amino acids, (B) arginine, (C) glycine, (D) threonine, (E) inosine, (F) lysine. Note that most values of lysine are zero in (F).

Table 2. Genome-wide significant single-nucleotide polymorphisms for the tested traits

Traits	SNP ID	Chr	Position (bp)	Allele 1	Allele 2	MAF	SNP effect	p-value
Essential FAAs	Gga_rs15943371	2	33,096,495	T	C	0.38	0.25	6.08×10^{-7}
Essential FAAs	Gga_rs14162297	2	33,432,575	T	C	0.42	0.24	7.12×10^{-7}
Essential FAAs	Gga_rs16609168	7	31,241,647	C	T	0.26	0.29	8.09×10^{-7}
Arginine	Gga_rs16609168	7	31,241,647	C	T	0.26	0.32	2.16×10^{-8}
Arginine	Gga_rs13971906	1	170,582,107	A	G	0.36	0.24	1.54×10^{-6}
Inosine	GGaluGA265714	4	74,314,272	G	A	0.39	-0.32	1.63×10^{-8}
Inosine	GGaluGA265712	4	74,285,145	A	G	0.39	-0.32	2.08×10^{-8}
Inosine	GGaluGA265969	4	75,827,200	T	C	0.30	-0.34	4.08×10^{-8}
Inosine	Gga_rs13971906	1	170,582,107	A	G	0.36	0.30	1.27×10^{-6}
Glycine	GGaluGA028030	1	81,831,988	T	C	0.06	0.16	9.37×10^{-8}
Threonine	Gga_rs14412837	3	108,537,571	G	A	0.50	-0.06	3.47×10^{-7}

Chr, chromosome; bp, base pair; Allele 1, minor allele; Allele2, major allele; MAF, minor allele frequency; SNP, single nucleotide polymorphism; FAA, free amino acids.

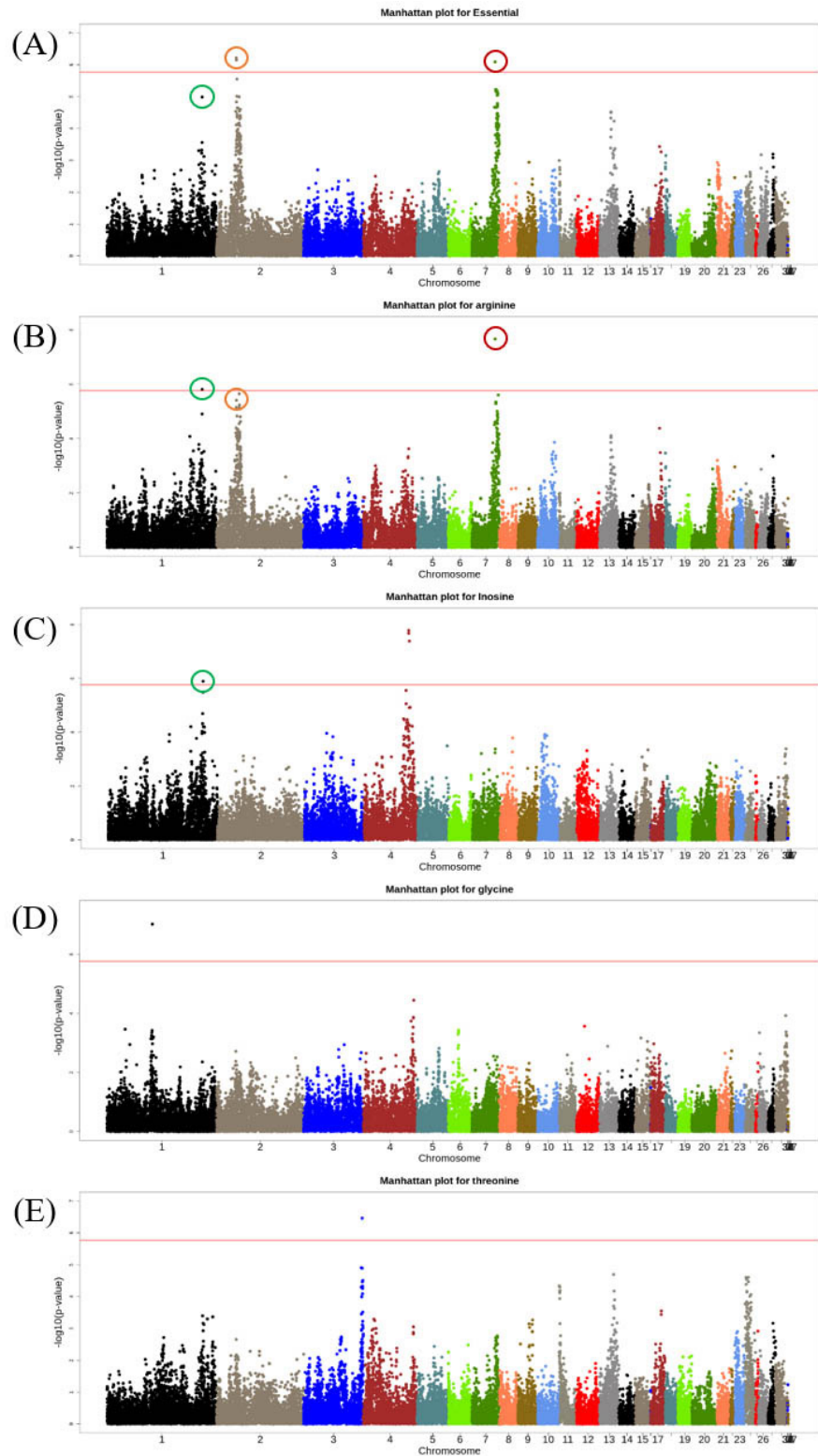


Fig. 3. Genome-wide Manhattan plot of the $-\text{Log}_{10}(p\text{-value})$ for significant single-nucleotide polymorphisms. (A) the essential free amino acids, (B) arginine, (C) inosine, (D) glycine, and (E) threonine components. The x-axis shows the chromosome number and the y-axis indicates the $-\text{Log}_{10}(p\text{-value})$. The red horizontal line indicates a Bonferroni-corrected significance level of 5%. Circles with the same color indicate the same genomic region.

SNP on GGA1 was rs13971906 for the essential FAAs, arginine, and inosine. Gga_rs14162297, which had the second highest significance for the essential FAAs, also had the second highest significance for arginine. Table 3 shows the significantly related genomic regions for each trait and the genes included therein.

DISCUSSION

The shared significant genomic region between the arginine and the essential free amino acids content

In this study, the candidate genes related to FAA and NT contents of chicken breast meat were identified through the GWAS analysis. In the GWAS result, the trend of significance for the essential FAAs was similar to that of arginine (Fig. 3). The reason why the two traits shared a significant genomic region in GGA1, GGA2, and GGA7 was that arginine content accounted for the largest portion of the essential FAAs in the phenotype. The arginine content in the breast meat of the F₂ crossbred chicken was greater than the content of all other FAAs when the mean values were compared.

Candidate genes in the shared significant region of the arginine and inosine contents

From the GWAS results of inosine and arginine, it was observed that the genomic region on GGA1 was commonly significantly related (Table 2). One of the genes included in this region, the *KCNRG* gene, encoded a voltage-gated potassium channel regulator protein. The voltage-gated potassium channel can regulate the nitric oxide synthesis pathway, which is a process in which L-arginine is converted to nitric oxide. This might have an effect on the arginine content in the body [20,21]. In addition, the voltage-gated potassium channel can modulate the insulin sensitivity of human skeletal muscle, which modulates the nitric oxide pathway and affects L-arginine transport and metabolism. Therefore, the potassium channel may be related to the amount of L-arginine in muscle [22–24]. Interestingly, if the content of adenosine triphosphate (ATP), a glucose metabolite in the muscle, changes as insulin manages skeletal muscle glucose uptake, the content of inosine, which is a repeatedly degraded form of ATP, may also be affected [25,26]. For the same reason, it could be concluded that the *KCNIP4* gene on the significant region in GGA4 encodes a potassium channel interacting protein.

Table 3. Genes located in the genomic regions for the significant single-nucleotide polymorphisms

Trait	Chr	Start position (bp)	End position (bp)	Genes
Arginine, Inosine	1	170,065,610	170,700,171	<i>RB1, LPAR6, RCBTB2, CYSLTR2, FNDC3A, MLNR, CDADC1, CA-B39L, SETDB2, PHF11, RCBTB1, ARL11, KPNA3, SPRYD7, TRIM13, KCNRG</i>
Arginine, Essential FAAs	2	32,588,887	33,940,447	<i>HOXA3, HOXA4, HOXA5, HOXA6, HOXA7, HOXA9, HOXA10, HOXA11, HOXA13, EVX1, HIBADH, TAX1BP1, JAZF1, CREB5, TRIL, CHN2, WIPF3, SCR11, FKBP14, PLEKHA8</i>
Arginine, Essential FAAs	7	30,761,616	31,457,806	<i>CXCR4, THSD7B, HNMT, SPOPL, NXPH2</i>
Inosine	4	74,767,885	75,920,712	<i>KCNIP4, ADGRA3, PPARGC1A, LDB2, QDPR, CLRN2, LAP3, MED28, FAM184B, LCORL, NCAPG</i>
Glycine	1	81,317,906	82,292,366	<i>CHD1L, FAAHL, TMEM39A, B4GALT4, B4GALT, B3GAT1L, UPK1B, IGSF11, LSAMP</i>
Threonine	3	108,221,189	108,846,159	<i>TFAP2B, TFAP2D, CRISP3, CRISP2, RHAG, CYP2AC1, CYP2AC2, CENPQ, MMUT, OPN5L2, CDC5L</i>

Chr, chromosome; bp, base pair; FAA, free amino acid.

Diet and nutritional level in birds are factors affecting the chemical composition of muscles. Among the candidate genes related to dry matter intake in cattle reported in Javaid et al. [27] and Brunet et al. [28], the *RB1*, *LPAR6*, *RCBTB2*, *CYSLTR2*, *FNDC3A*, *MLNR*, *PHF11*, and *KPNA3* genes were identical to the genes in the genomic region commonly related to inosine and arginine in this study. Therefore, it is possible that the difference in nutritional condition according to the feed intake of chickens raised under *ad libitum* conditions in this study affected the chemical composition of breast meat [27].

Candidate genes related to arginine and threonine

It has been reported that the arginine content can be increased by reducing nitric oxide synthase when the expression of the *HOXA3* gene among the *GGA2* genes related to arginine content in Table 3 is regulated [29]. It has also been confirmed that mutations in the *THSD7B* gene of *GGA7* affect nitric oxide metabolism [30]. In the GWAS result of the threonine content, the *MMUT* gene encoding the methylmalonyl-CoA mutase located on *GGA3* was included as a candidate gene. In animals, including chickens, threonine is catabolized to alpha-ketobutyrate by serine/threonine dehydratase, then converted to propionyl-CoA and carboxylated to methylmalonyl-CoA [31–33]. The catalyzed product is finally converted to succinyl-CoA by methylmalonyl-CoA mutase and used in the tricarboxylic acid cycle [34]. Therefore, threonine content may be affected by methylmalonyl-CoA mutase involved pathway which use threonine as a substrate.

Further improvements from the present study

The molecules with significantly related SNP markers in this study were arginine, inosine, threonine, and glycine. As taste-active precursors that determine meat flavor during cooking, arginine and inosine give a bitter taste, and threonine and glycine give a sweet taste [9]. This study had limited success in finding a significant gene region related to the contents of FAA and NT molecules that give umami and sour tastes in chicken breast. The reason that significant SNPs were not found through GWAS in most traits is thought to be due to the large impact of environmental effects on the FAA and NT contents. The FAA content increases as protease hydrolyzes meat protein, and this degradation proceeds even in frozen meat [35]. A significant change in the amino acid profile of meat according to the storage duration at freezing temperature has also been reported [36]. In addition, in the process of thawing frozen meat, several NT-metabolizing enzymes perform a chain degradation to change the NT content [37–39]. Based on the results of this study, it could be concluded that the period of storage under freezing conditions before analysis after the slaughter of chickens was long enough to result in significant changes in the FAA and NT contents. Future studies may capture more genetic effects when association tests are conducted by performing phenotyping on meat with a shorter storage time than in the present study.

CONCLUSION

Changes in the concentrations of FAAs and NTs in chicken are the main factors that change the flavor. To our knowledge, this was the first GWAS to explore genetic markers related to the content of FAAs and NTs in chicken. Through the GWAS, nine significant SNPs related to essential FAAs, arginine, glycine, threonine, and inosine components were found, and six significant genomic regions and candidate genes included therein were reported. These genomic regions were mapped on chromosome 1, 4, and 7 of chicken, and the potential candidate genes were *KCNRG* for arginine and inosine, *HOXA3* and *THSD7B* for arginine, *KCNIP4* for inosine, and *MMUT* for threonine. The reported candidate genes might be important genetic markers in chicken selection

and breeding to help meet consumer expectations for healthier, tastier, and flavored meat.

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