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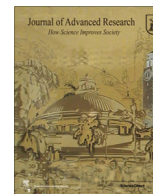
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Original Article

Whole genome analyses reveal novel genes associated with chicken adaptation to tropical and frigid environments



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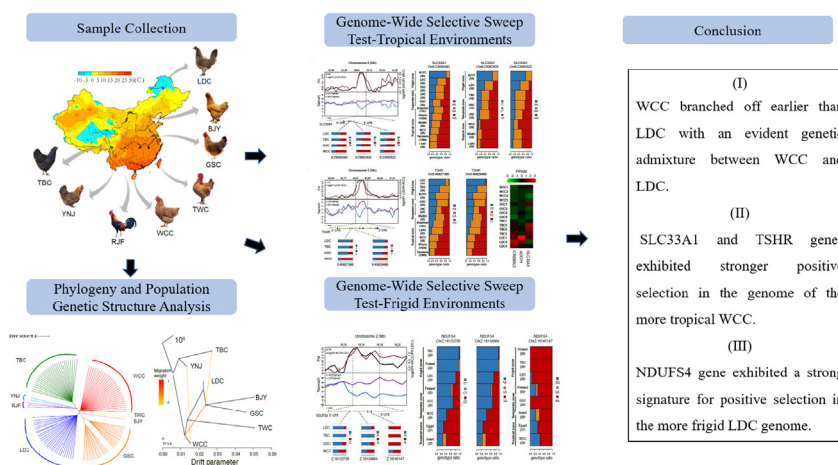
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HIGHLIGHTS

- Chickens from Southern China originated earlier than those from Northern China.
- SLC33A1 and TSHR genes provided key evolutionary molecular adaption to tropical environment.
- 5 SNPs in SLC33A1 and TSHR appeared low mutant allele frequencies to adapt to tropical environment.
- NDUFS4 gene helped chicken achieve rapid adaption to frigid environment.
- SLC33A1, TSHR and NDUFS4 can be biomarkers to assess the adaptation to extreme environments.

GRAPHICAL ABSTRACT



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ABSTRACT

Introduction: Investigating the genetic footprints of historical temperature selection can get insights to the local adaptation and feasible influences of climate change on long-term population dynamics.

Object: Chicken is a significant species to study genetic adaptation on account of its similar domestication track related to human activity with the most diversified varieties. Yet, few studies have demonstrated the genetic signatures of its adaptation to naturally tropical and frigid environments.

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Keywords:

Chicken
 Extreme environment
 Whole-genome sequence
 Genetic adaption

Method: Here, we generated whole genome resequencing of 119 domesticated chickens in China including the following breeds which are in order of breeding environmental temperature from more tropical to more frigid: Wenchang chicken (WCC), green-shell chicken (GSC), Tibetan chicken (TBC), and Lindian chicken (LDC).

Results: Our results showed WCC branched off earlier than LDC with an evident genetic admixture between WCC and LDC, suggesting their closer genetic relationship. Further comparative genomic analyses of solute carrier family 33 member 1 (*SLC33A1*) and thyroid stimulating hormone receptor (*TSHR*) genes exhibited stronger signatures for positive selection in the genome of the more tropical WCC. Furthermore, genotype data from about 3,000 African local ecotypes confirmed that allele frequencies of single nucleotide polymorphisms (SNPs) in these 2 genes appeared strongly associated with tropical environment adaptation. In addition, the NADH:ubiquinone oxidoreductase subunit S4 (*NDUFS4*) gene exhibited a strong signature for positive selection in the LDC genome, and SNPs with marked allele frequency differences indicated a significant relationship with frigid environment adaptation.

Conclusion: Our findings partially clarify how selection footprints from environmental temperature stress can lead to advantageous genomic adaptations to tropical and frigid environments in poultry and provide a valuable resource for selective breeding of chickens.

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Introduction

Environment pressures have the feasibility to adversely affect the health of domestic animals, with consequences to reducing animal welfare and economic efficiency [1]. On the strength of differences of evolved genetic adaptations, resilience obtained from artificial and natural selection showed significant differences among varieties [2], which has resulted in a wide diversity of domestic animal breeds since they were domesticated. Hence, domestic animals are excellent models for genetic studies and identification of mutations underlying evolution adaption [3], which have contributed to basic and medical biology [4]. Due to the convenience and simple requirements of feeding conditions during migration, domestic chicken is closely related to human activity tracks and has become the most diversified domestication model among domestic animals [5]. According to 2007 FAO statistics, China owned 116 domestic chicken breeds, accounting for nearly 10% of the world's breeds [6]. China situated in eastern Asia, and is located on the west coast of the Pacific Ocean, and it straddles the eastern and northern hemispheres with diversified geographical, climatic and ecological conditions. In particular, there is a great temperature difference (~76 °C between hottest and coldest temperature) from North to South created the conditions for the formation of variety diversity and evolved genetic adaptations in chickens.

Considering their short breeding and growth periods, and a similar domestication track related to human activity with the most diversified domestic animals, chickens have served as model organisms for investigating genetic diversity and evolutionary selection mechanisms [7,8]. Several studies have reported that the domestication of chicken started in multiple regions including South Asia, Southeast Asia, and Southwestern China thousands of years ago [9]. Subsequently, numerous indigenous local breeds were formed after a long time of natural and artificial selections, which have adapted to extensive eco-geographic conditions, especially those in tropical and frigid regions. In view of the current and future global influences of climate change, understanding the genetic adaptation of local chicken breeds to extreme environments and then developing applicable breeding plans to tackle climate change are of practical value for the modern chicken industry.

Extreme environment pressures have been considered as an important driver shaping the animal genome [10]. Identifying an adaptive selection signature driven by extreme environments has become a key focus of evolutionary biology. A series of genetic variations influenced by environmental adaptation have been reported in diverse species, such as *Drosophila*, chickens, honey bees, sheep, and horses [11–15]. Exploring the selective signature

mediated by climate necessary for comprehending the genetic basis of native environmental adaption as well as any functionally vital variants [16]. For the past few years, whole genomic sequencing (WGS) has been widely used for genomic studies in domestic livestock. In fact, WGS can identify candidate genes of important traits and utilize SNPs to reveal detailed insights into biological questions such as evolutionary processes, environmental adaptation, and natural selection [17]. Previous genomic studies on poultry have mainly been conducted on the genetic diversity of breeds [18], evolutionary mechanisms [19], and genetic mechanisms of quality traits [20]. However, to our knowledge, there is a paucity of WGS studies characterizing poultry genetic adaptations to extreme environments, like tropical and frigid environments, although there has been some genetic work on adaptation of Tibetan chickens (TBC) to high altitudes [12] and tropical desert climate [7].

In this study, the whole genomes of 119 chickens from 4 Chinese local breeds in various temperature environments were sequenced (Fig. 1A and supplementary Fig. S2B, supplementary table S1, Supplementary Material online). Then we performed comprehensive analyses of the population structure, genomic diversity, and demographic history of those chicken breeds according to their genomic data. We also identified candidate genes and pathways that were potentially associated with rapid adaptation to tropical and frigid environments, since domestication of the chicken occurred in Southern Asia ~ 6,000–8,000 years ago [21].

Materials and methods

Ethics statement

Handling and sampling of 119 Chinese local chickens were conducted under the guidance of ethical regulations of the Poultry Institute, Chinese Academy of Agriculture Science, Jiangsu, China. All experiments involving chickens from Tanzania and Ghana were conducted according to the ethical policies and procedures of the Institutional Animal Care and Use Committee, the University of California, Davis (#:17853).

Sample collection

Blood was collected from 119 chickens for whole-genome sequencing, including 29 Wenchang chickens (female, 43 weeks, Wenchang, Hainan, 32 °C), 30 Green-shell laying hens (female, 42 weeks, Yangzhou, Jiangsu, 20 °C), 30 Lindian chickens (female, 42 weeks, Lindian County, Heilongjiang, –3 °C) and 30 Tibetan chickens (female, 40 weeks, Ganzi, Sichuan, 2 °C) from their national

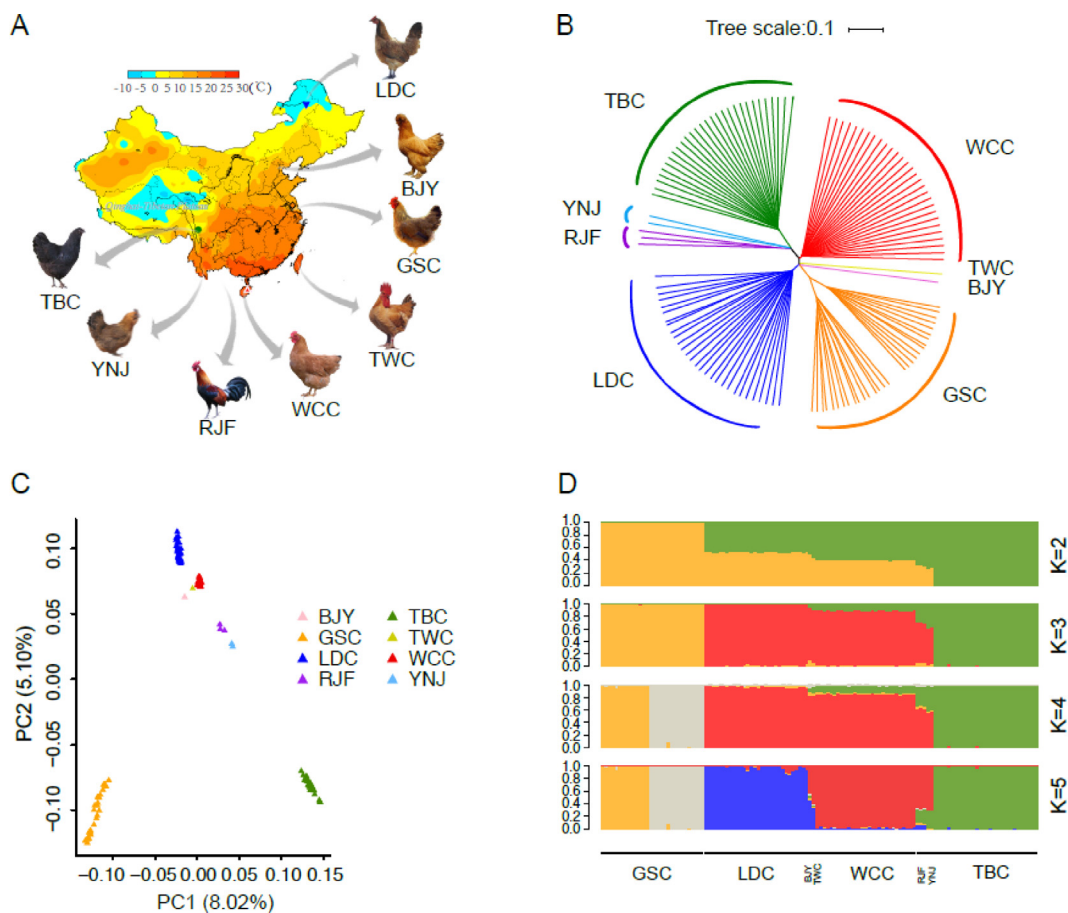


Fig. 1. Population genetics. (A) Geographic locations of the 126 individuals. A total of 119 sampled Chinese chickens representing four breeds were included and 7 individuals of four breeds were used as outgroups. (B) The neighbor-joining tree constructed using *p*-distance between individuals. The 7 individuals of four breeds were used as outgroups. (C) Principal component plot of the 126 individuals. (D) Population genetic structure of the 126 individuals examined via the program FRAPPE v1.1. The number of assumed genetic clusters *K* ranged from 2 to 5 are shown.

local poultry genetic resources conservation farm with different temperature zones (supplementary table S1, [supplementary Fig S2B](#), [Supplementary Material](#) online). After blood collection of these 119 chickens, 16 liver tissues (4 chickens per breed) were sampled for RNA extraction and transcriptome analyses. Also, blood samples of 120 chickens originating from four local breeds (30 individuals per breed), including Mingqing Mao Jiao chicken from Fujian (female, 43 weeks, 30 °C), Longsheng Feng chicken from Guangxi (female, 41 weeks, 29 °C), Jining Baire chicken from Shandong (female, 46 weeks, 17 °C), and Beijing You chicken from Beijing (female, 44 weeks, 13 °C), were collected for DNA extraction and genotyped by PCR. All these birds were reared in semi open chicken house and fed with traditional corn-soybean meal diet.

Blood samples of 1,399 indigenous chickens from the regions of Morogoro, Tanga, Shinyanga, Mwaza, and Singida in Tanzania (29 °C), as well as 1,439 indigenous chickens from Ghana (30 °C) were obtained for chicken Affymetrix 600 K SNP chip analysis. All the blood samples were collected from wing vein and stored in 5 ml EDTA sodium anticoagulant tube at -20 °C.

Whole genome sequencing library preparation

The DNA was extracted with 1.5 µg per sample. We formed sequencing libraries via Truseq Nano DNA HT Sample preparation Kit (Illumina USA) and then added index codes to each sample. In brief, every DNA sample was fragmented into 350 bp size, which were then end polished, A-tailed, and ligated. At last, we purified

PCR products and analyzed these libraries by Agilent2100 Bioanalyzer as well as quantified using real-time PCR.

Genome sequencing

In the current research, we sequenced the whole genome of these samples with the Illumina HiSeq 2000 platform. In all, 3287.41 Gb raw data were sequenced. Firstly, low quality paired reads were removed (Reads with ≥ 10% unidentified nucleotides (N); > 10 nt aligned to the adaptor, allowing ≤ 10% mismatches; > 50% bases having phred quality < 5; and putative PCR duplicates generated in the library construction process). Consequently, we generated 3278.18 Gb (~22.74-average fold per sample) high quality paired-end reads for 119 samples with the quality of 93.44% and 90.75% of the bases were ≥ Q20 and ≥ Q30, respectively.

Reads mapping

The rest of high quality reads were mapped to the Gallus_gallus-5.0 reference genome using BWA (Burrows-Wheeler Aligner) (Version: 0.7.8) with the command 'mem -t 4 -k 32 -M' [22]. In order to decrease mismatch generated by PCR amplification, duplicated reads were removed by Samtools v1.3.1 [23].

SNP and InDel calling

SNP and InDel were performed by a Bayesian approach which was carried out with the package Samtoolsv1.3.1 [24]. Then we calculated the genotype likelihoods from reads at each genomic location, and allele frequencies were calculated with a Bayesian approach. The 'mpileup' command was applied to authenticate SNPs with the parameters like '-q 1 -C 50 -t SP -t DP -m 2 -F 0.002'. Then, only high quality SNPs (coverage depth ≥ 5 , RMS mapping quality ≥ 20 , maf ≥ 0.05 , miss ≤ 0.1) were reserved for pursuant analysis to eliminate SNP calling errors caused by incorrect mapping or InDels. Hence, 11,690,980 SNPs were reserved after filtering from 22,830,447 raw SNPs.

Variants of structure detection

Structure variants (SV) including insertions (INS), deletions (DEL), intra-chromosomal translocations (ITX), inter-chromosomal translocations (CTX) and inversions (INV) of all 119 samples and 4 groups was detected by lumpy-sv v 0.2.13 [25].

Functional annotation of genetic variants

We annotate those SNP and InDel according to the Gallus_gallus-5.0 genome using the package ANNOVAR [24] (Version: 2013–05-20). SNPs were grouped into intronic regions, exonic regions, upstream and downstream regions, splicing sites, and intergenic regions according to the genome annotation results. Then those SNPs in coding exons were moreover classed into synonymous SNPs or nonsynonymous SNPs.

Phylogenetic tree and population structure

We instituted a neighbour-joining (NJ) tree by the software TreeBest v1.9.2 to show the phylogenetic relationship based on the genome-wide SNPs. An expectation maximization algorithm was applied to inspect the population genetic structure in the program FRAPPE v1.1. We made the Principal component analysis (PCA) via the software GCTA v1.24.2. Then Tracey-Widom experiment was used to evaluate the eigenvector significance level.

Genomic diversity and linkage disequilibrium

The $\theta\pi$ was calculated by Vcftools v0.1.14, and θw was calculated by variscan 2.0. The ROH (regions of homozygosity) was identified by the runs of homozygosity tool in PLINK (v.1.07) with the parameters (--homozyg-density 10 --homozyg-window-het 1 --homozyg-window-kb 5000 --homozyg-window-snp 20 --homozyg-kb 100 --homozyg-snp 10). Then we conducted the squared correlation coefficient (r^2) with the software Haploview [26] to estimate LD decay. The average r^2 was counted in a 500-kb window as well as averaged through the whole genome.

Population Historical, differentiation time and migration events

Demographic history of 119 samples was reconstructed by a hidden Markov model (HMM) approach in pairwise sequentially Markovian coalescent (PSMC v 0.6.4-r49) [27]. Initially, genotype of each sample was called via the package Samtools v1.3.1. Then, the program 'fq2psmcfa' was applied to switch the consensus sequence into a fasta-like format where the i-th character indicates whether there is at least one heterozygote in the bin [100i, 100i + 100], which the purpose of improving the accuracy of inferred historical recombination events. The parameters were set as follows: -N30, -t15, -r5 and -p '4 + 25*2 + 4 + 6'. The

chicken neutral mutation rate per generation (μ) was 0.19×10^{-8} and generation time (g) was 1 years [12].

We estimated the divergent time by fastsimcoal2 v 2.6.0.3 [28], SNP located at intergenic regions were applied to transform SFS by easySFS (<https://github.com/isaacovercast/easySFS#easysfs>). Mutation ratio was set to $2E-9$ per site per generation as well as generation time as 1 year. The fastsimcoal2 was conducted 100 times with changed starting points, and the fitting model was reserved based on the highest likelihood. Demographic estimates were obtained from 100,000 simulations (-n100,000) per parameter file, 40 Expectation/Conditional Maximization cycle (-L40). The population relationship and gene flow patterns were concluded with a maximum likelihood approach applied in TreeMix v 1.12 [29].

Genome-wide selective sweep test

To ascertain genome-wide selective sweeps concerned with temperature adaptation, the genome-wide assignment of fixation index (F_{ST}) values and $\theta\pi$ ratios were calculated for these defined group pairs (vcftools v0.1.14). Average SNP F_{ST} values were plotted in 40 kb genomic bins with a 20 kb step. Nucleotide diversity (π) was estimated for the same bins. The F_{ST} values were Z-transformed as follows: $Z(F_{ST}) = (F_{ST} - \mu F_{ST}) / \sigma F_{ST}$, in which μF_{ST} is the mean F_{ST} and σF_{ST} is the standard deviation of F_{ST} . The $\theta\pi$ ratios were log2-transformed. Then we evaluated and arranged the empirical percentiles of $Z(F_{ST})$ and $\log_2(\theta\pi$ ratio). The windows with the top 5% $Z(F_{ST})$ and $\log_2(\theta\pi$ ratio) values contemporaneously considered as nominated outliers under strong selective sweeps. XP-EHH value of each SNP was calculated using the XP-EHH program (<https://github.com/joepickrell/xpehh>). The average XP-EHH values were computed with a 40 kb window size and a 20 kb step size. Windows with the top 5% XP-EHH values were selected.

Genotyping array analysis

The blood samples of chickens from Tanzania and Ghana were applied for DNA extraction and Array analysis. Genomic DNA was isolated and genotyping was analysed via the Aymetrix Axiom[®] 600 k Array at GeneSeek (Lincoln, NE, USA). Annotation files of Genotyping Array were referred to the Gallus gallus genome version 5 (Thermo Fisher Scientific Inc., Calsbad, CA, USA). We then performed the genotype data quality filtering by AxiomTM Analysis Suite 3.1 (Applied Biosystems, Thermo Fisher Scientific Inc., Calsbad, CA, USA) and these genotype data quality filtering included single nucleotide polymorphism (SNP) call rate $\geq 99\%$ as well as minor allele frequency ≥ 0.05 .

Transcriptome sequencing and analysis

Liver tissues from 16 chickens (4 biological replicates per condition) were collected and Sequenced for transcriptomic analysis. We generated sequencing libraries via NEBNext[®] UltraTM RNA Library Prep Kit for Illumina[®] (NEB, USA). Then we conducted the clustering of the index-coded samples on a cBot Cluster Generation System using TruSeq PE Cluster Kit v3-cBot-HS (Illumia). Then the library preparations were sequenced via an Illumina Hiseq platform and 150 bp paired-end reads were created. Totally, 126.86 Gb RNA-seq reads were generated with 16 liver tissues. Index of the reference genome was built with Bowtie v2.2.3 and aligned the high-quality RNA-seq reads to the reference genome by the HISAT2 v2.0.4 program. HTSeq v0.6.1 was applied to cast the reads numbers of each gene. Then we calculated FPKM according to gene length and read count numbers.

Quantitative PCR analysis and Western blotting

qPCR mainly was conducted for five genes, *SLC33A1*, *NFUFS4*, *TSHR* and its downstream genes adenylate cyclase 1 (*ADCY1*), cAMP responsive element binding protein 3 like 3 (*CREB3L3*). The primer sequences of these genes were revealed in Supplementary Table S23, [Supplementary Material](#) online. Liver tissues randomly selected from 5 chickens per breed were used to verify the accuracy of the transcriptome data, and those heart and hypothalamus tissues were performed to reveal the temporal expression pattern. Total RNA was isolated and synthesized cDNA using FastKing Super Mix Kit (KR118, Tiangen Biotech Co., Ltd.). Then, we used SuperReal PreMix Plus (FP205, Tiangen Biotechnology Co., Ltd.) to perform Real-time PCR by StepOnePlus (Applied Biosystems). And the mRNA expression was calculated by $2^{-\Delta\Delta Ct}$ method.

20 Liver proteins (5 chickens per breed) were extracted using RIPA Lysis Buffer. After complete lysis of samples, we collected the supernatant, and tested the protein concentrations with a BCA Protein Assay kit. Each sample was denatured at 100 °C for 5 min before being loaded into polyacrylamide gel and then moved to polyvinylidene fluoride membranes. The membranes were washed in Tween-Tris-buffered saline buffer and blocked for 2 h, and then cultivated by primary antibodies (*SLC33A1* (LS-B11505, Lifespan) at 1:500, *TSHR* (PAB27584, abnova) at 1:500, *NDUFS4* (DF9670, affinity) at 1:500, β -actin at 1:2000) at 4 °C overnight. After washes, we incubated the membranes via anti-rabbit IgG secondary antibody for 2 h. Then we visualized the immunoblots with enhanced chemiluminescence detection and quantified via densitometric scanning of the autoradiograph. The band densities of proteins were normalized with those of β -actin.

Results

Sequencing variants among different chicken breeds

We conducted whole-genome resequencing of 119 individuals from the 4 local chicken breeds which are in order of breeding environmental temperature from more tropical to more frigid: Wenchang chicken (WCC), Green-shell chicken (GSC), TBC, and Lincian chicken (LDC). We acquired a total of 3,278 Gb of high-quality genomic data, with an average of 183.61 M paired-end reads (150 bp) per individual. Alignment of the obtained sequences to the chicken reference genome (ftp://ftp.ensembl.org/pub/release-90/fasta/gallus_gallus/dna/Gallus_gallus.Gallus_gallus-5.0.dna.top.level.fa.gz) resulted in an average of 92.52% sequencing coverage with $22.74 \times$ effective depth of each individual ([supplementary table S2, S3, Supplementary Material](#) online). After accurate quality filtering of the identified SNPs, 11,690,980 high-quality SNPs were identified in the 119 individuals. Of these SNPs, 83.55% ~ 88.81% were found in the chicken dbSNP database ([supplementary Fig. S2A, supplementary table S4, Supplementary Material](#) online), proving the high quality and validity of the SNP calling. Of these four populations, a majority of these SNPs identified in the chicken genome were lied in intronic and intergenic regions (11.11 million, 95.03%), while only 1.41% (0.17 million) of the SNPs were in exonic regions ([supplementary table S5, Supplementary Material](#) online). And 117,539 synonymous SNPs and 47,151 nonsynonymous SNPs were located in exons, resulting in a nonsynonymous/synonymous ratio of 0.40, which was consistent with previous studies in chickens [18]. In addition, we detected a total of 2,212,046 indels (<100 bp), with 1,942,963 for the WCC, 1,530,616 for the GSC, 1,757,682 for the TBC, and 1,780,716 for the LDC ([supplementary table S6, Supplementary Material](#) online). In total, 18,315 structural variations (SVs, 100 bp-chromosome level) were identified including deletion, inversion, and complex rearrangements with

breakends ([supplementary table S7, Supplementary Material](#) online). Most of the indels and SVs were in intergenic regions ([supplementary table S6, supplementary table S8, Supplementary Material](#) online). And all the SNPs, indels and SVs distributed evenly on all chromosomes ([supplementary Fig. S1, Supplementary Material](#) online).

Phylogeny and population genetic structure as well as migration events

To infer genetic and evolutionary relationships of chicken breeds adapted to diverse temperature environments, we generated high quality whole genome sequencing data of 119 individuals from 4 populations, as previously mentioned. In addition, we also downloaded resequencing data from the NCBI database for the following four chicken breeds: Red jungle fowl (RJF, 3 individuals), Yunnan native chickens (YNC, 2 individuals), Taiwan chickens (TWC, 1 individual), and Beijing-You chickens (BJY, 1 individual) ([Fig. 1A, supplementary table S9, Supplementary Material](#) online). From these data we in all obtained 11,698,251 SNPs. Next, a phylogenetic tree from a total of eight chicken breeds with 126 chickens was constructed using the neighbor-joining (NJ) algorithm according to pairwise genetic distances between populations according to the polymorphic SNPs in the whole genome. The tree revealed that individuals from the same geographic region tended to cluster into a single group ([Fig. 1B](#)). RJF, as the ancestor of the domestic chicken, was in the root of the tree and clustered together with TBC and YNJ, which were 2 ancient native breeds raised in Southwest China near the geographic region of RJF. Those 3 breeds first branched off, then WCC, TWC, BJY, LDC and GSC successively branched off in order of their geographic position from south to north and environment from tropical to frigid. GSC branched off relatively late as it was a commercial breed selected by artificial selection. These findings were corroborated by both principal component analysis (PCA) and population structure analysis ([Fig. 1C and 1D, supplementary Fig. S4, Supplementary Material](#) online).

In the population structure analysis, when $K \leq 3$, the overall individuals were genetically fallen into three clusters (WCC-LDC, GSC, or TBC populations) with WCC and LDC clustering together and the other 2 populations tending to cluster independently. When $K = 4$, two genetic clusters were formed within the population of GSC. This was in accordance with NJ tree and PCA analysis and revealed an obvious branch in the GSC population, suggesting that there are 2 different strains of GSC. As a commercial high-yield breed, plumage color change in GSC (black to hemp-colored feathers) appeared during the hybridization process. Also, we detected a number of nonsynonymous mutations in several plumage color related key genes including melanocortin 1 receptor (*MC1R*) and premelanosome protein (*PMEI*) [30,31], with different SNP frequencies in the 2 GSC strains ([supplementary Table S10, Supplementary Material](#) online).

Furthermore, we observed a clear signature of genetic introgression or ancestral polymorphisms from TBC into WCC and LDC when $K = 2$. We also found WCC and LDC clustered more closely with each other from $K = 2$ to $K = 4$, and there was a slight gene flow between them ([Fig. 1D](#)). A division between WCC and LDC was found at $K = 5$, which was the optimal K , with slight genetic admixture between these two breeds, indicating a close genetic relationship and a similar domestication history between the two breeds, although the precise origins of WCC and LDC are unclear. Demographic modeling using Fastsimcoal analysis supported the idea that LDC originated from breeds of Southern China ([Fig. 2B](#)), further confirming the close relationship between WCC and LDC.

We then rebuilt the maximum likelihood tree and residual matrix of these breeds and added sequential migration events to the tree by using TreeMix [29], to address population history rela-

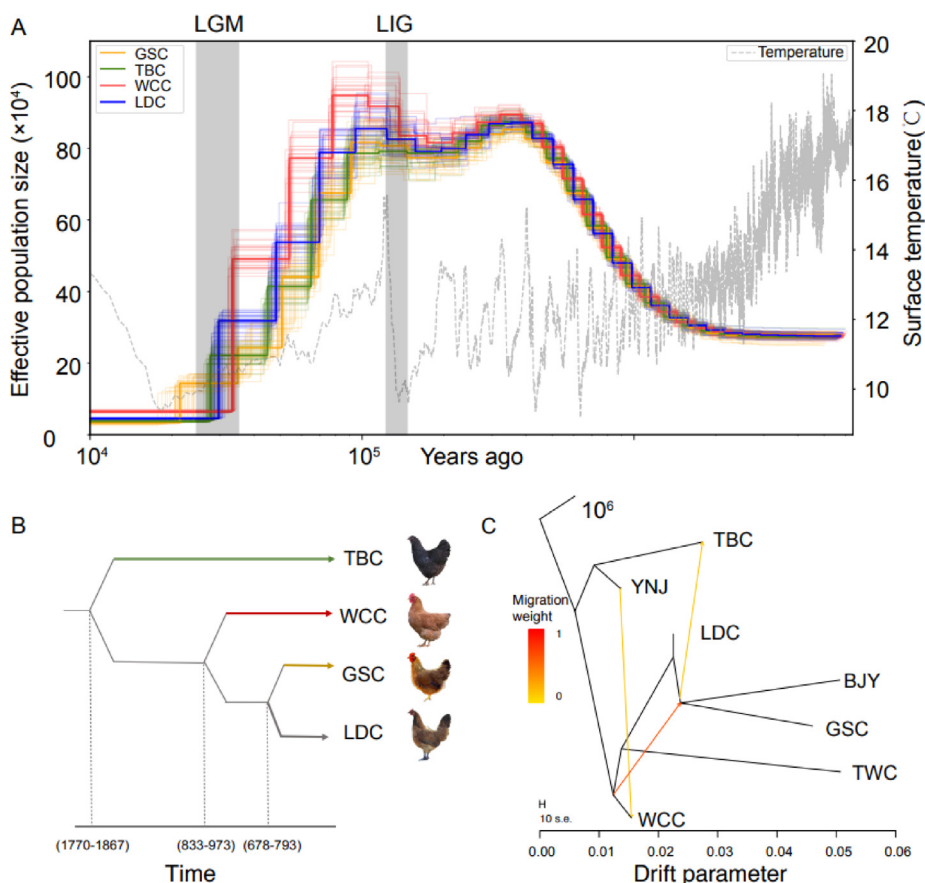


Fig. 2. Demographic history of the chicken population. (A) Estimated effective population size of the 119 individuals with PSMC analysis based on SNP distribution. (B) The demographic history of the four populations from ~ 2,000 years ago to present showed with fastsimcoal analysis. (C) Pattern of population splits and mixtures among the four populations and the four outgroups. The drift parameter is proportional to N_e generations, where N_e is the effective population size. Scale bar shows ten times the average standard error of the estimated entries in the sample covariance matrix. Arrows indicate migration events, and a spectrum of heat colors indicate the migration weights of the migration events.

tionships and to deduce migration patterns among those different breeds (Fig. 2C, supplementary Fig. S6, Supplementary Material online). We found that one supposed migration edge produced a tree with the smallest residuals and consequently best fit the data. We observed a statistically significant migration edge ($p < 2.23 \times 10^{-308}$) with an estimated greatest weight of 62.65%. This edge provides evidence for the gene flow from WCC to the breeds in Northern China. In recent years, studies have provided a perspective on the pattern of gene flow between Southeastern and Northern breeds in China [32], though no direct evidence has been presented to demonstrate that WCC has been crossbred with Northern breeds. Similar to the previous TreeMix finding [33], our data found gene flow between LDC and TBC. Also consistent with previous results [34], current study revealed gene flow between YNC and WCC, which might be attributable to population migration from the Southwest to Hainan China [35]. These findings partly provided confidence to a single origin theory for Southern and Northern chicken breeds and frequent gene flows among them, further implying the multiple and close relationships of these breeds.

Genomic diversity and linkage disequilibrium

The genome-wide average $\theta\pi$ values of the four populations were estimated in this study, with the highest $\theta\pi$ value (2.99×10^{-3}) being observed in WCC while the lowest $\theta\pi$ value (2.64×10^{-3}) was found for GSC. The other genomic variation

parameters (average number of regions of homozygosity (ROH) and mean ROH size) showed nearly congruent patterns (supplementary Fig. S2C - S2F, supplementary table S11, Supplementary Material online). In other words, WCC showed the lowest ROH number (377) and a relatively smaller average ROH size (347.87 kb), whereas the highest ROH number (581) and the largest average ROH size (511.75 kb) were observed in GSC. These data further indicate the greatest genomic diversity in WCC and extensive inbreeding in GSC, potentially due to artificial selection. Genome-wide linkage disequilibrium of the four populations also varied markedly (supplementary Fig. S3, Supplementary Material online). Specifically, GSC showed a slow decay rate and a high level of linkage disequilibrium, while WCC exhibited a rapid decay rate and a low level of linkage disequilibrium. TBC and LDC presented a similar decay rate and level of linkage disequilibrium in an intermediate status. Taken together, the patterns of linkage disequilibrium and genomic diversity may reflect the differences in population histories of genetic bases, and the low genetic diversity of GSC may have resulted from artificial selection and breeding in recent years.

Population history and differentiation time

To explore the demographic history of the four chicken populations, the historical effective population sizes (N_e) of the chicken ancestors were estimated using PSMC method according to SNPs in the whole genome and the linkage disequilibrium among them.

The results revealed that individuals from the same populations or the same geographic regions exhibited a highly similar pattern of demographic trajectories (Fig. 2A, supplementary Fig. S5, Supplementary Material online). The N_e of the four populations fluctuated over time with the earth's climate, with two expansions and two bottlenecks in the demographic history, which is consistent with the findings of others [12]. We found that an increase in population sizes of the four chicken breeds occurred sometime around 1 Mya and reached a peak at approximately 100 kya, with a dramatic decline following that peak, which mirrors environmental changes in Southeast Asia where RJF are distributed. In time of the Quaternary ice ages (from ~ 2.5 Ma to now), the rainforests in Southeast Asia that is known as one of the world's biodiversity hotspots with a high concentration of local species, served as a refugia, empowering the persistence of forest biota [36]. Meanwhile a slight population size fluctuation occurred following that peak. This fluctuation was consistent with the Last Interglacial (LIG), during which increased seasonality in temperature extremes appeared [37]. An apparent reduction in population sizes of the four breeds was observed from the LIG until the Last Glacial Maximum (LGM), which might have resulted from a substantial decrease of the forested area that occurred in Southeast Asia during the Last Glacial (LG) period (~125–10 kya) [38]. In addition, it was clear that similar N_e s were observed in all 4 breeds, which implied none of the 4 breeds had branched off 10 kya ago. Fastsimcoal analysis showed TBC was likely domesticated more than 1,000 years, which was in accordance with a previous study [39]. WCC and LDC branched off ~ 800 and ~ 600 ya, respectively, indicating their time of domestication (Fig. 2B). Interestingly, we found the age estimate of LDC by fastsimcoal analysis coincided with the introduction time of WCC to Northern China. In fact, a historical record in "Lingnan omnivorous poem" [40], indicated that WCC was used as a tribute to the imperial court in the Ming Dynasty (CE 1368–1644). WCC was received by the emperor, and then WCC was introduced to Northern China and domesticated locally.

Genome-Wide selective sweep test

Adaptive mechanisms in tropical environments

To study adaptive mechanisms of chickens in tropical environments, we used comparisons between WCC and LDC to screen WCC for a selective signal (supplementary Fig. S7, Supplementary Material online). Next, using the top 5% of both F_{ST} and $\theta\pi$ ratio cutoffs, we identified 17.92 Mb of 448 selective sweep regions encompassing 269 genes (Fig. 3, supplementary table S12, Supplementary Material online) and several highly enriched pathways (supplementary table S15, Supplementary Material online), which were mainly associated with glycometabolism and energy metabolism. In the negative control group GSC vs LDC and TBC vs LDC, we discovered a 36.96 Mb of 924 selective sweep region encompassing 455 genes with three highly enriched pathways (supplementary Fig. S8, supplementary table S13, Supplementary Material online) and a 22.60 Mb of 565 selective sweep region encompassing 278 genes with 4 highly enriched pathways (supplementary Fig. S9, supplementary table S14, Supplementary Material online). In total, 5 pathways with ten genes in WCC had no overlap with the other seven pathways, and 5 genes including *solute carrier family 33 member 1* (*SLC33A1*); *golgi associated, gamma adaptin ear containing, ARF binding protein 3* (*GGA3*); *alpha-L-fucosidase 1* (*FUCA1*); *UDP-galactose-4-epimerase* (*GALE*); and *palmitoyl-protein thioesterase 1* (*PPT1*) were also identified using the XP-EHH method (Fig. 3), which further speculated their significant role in tropical environmental adaption.

Among the 10 candidate genes for tropical environments, three genes [*aldo-keto reductase family 1 member B10* (*AKR1B10L2*), *GALE*, and *galactose-1-phosphate uridylyltransferase* (*GALT*)] are in the

galactose metabolism pathway, and three genes [*GALE*, *hexosaminidase subunit beta* (*HEXB*), and *GALT*] are in the amino sugar and nucleotide sugar metabolism pathway (supplementary table S15, Supplementary Material online). Both are found upstream of the pentose phosphate pathway which plays an important role in high temperature conditions because it generates NADPH to serve as a cofactor of GP_x [41]. Meanwhile, UDP-Glc was found as an important metabolite in the amino sugar and nucleotide sugar metabolism pathway (Fig. 4A), which also participates in extreme environment adaption [42]. In addition, we discovered a positive metabolite, GlcNAc, which appears in the amino sugar and nucleotide sugar metabolism pathway and another glycan degradation pathway. It is a posttranslational sugar modification of nucleocytoplasmic and mitochondrial proteins which confers heat stress tolerance [43]. In the lysosome pathway, we also found 5 positively selected genes (*ATP6AP1L*, *HEXB*, *AP1G1*, *PPT1* and *GGA3*), which directly affect substances for digestion via a series of processes including endocytosis, phagocytosis, and autophagy and which ultimately participate in heat stress-induced intestinal tissue and epithelial cell apoptosis [44]. 2 genes (*HEXB* and *FUCA1*) affect another glycan degradation pathway. 2 genes (*HEXB* and *SLC33A1*) influence the glycosphingolipid biosynthesis-ganglio series, and they are functionally related to cell proliferation, differentiation, and adhesion [45]. These 5 relevant pathways encompassing glycometabolism and energy metabolism, which are related to cellular metabolism, are involved in responses to high-temperature stress [46] and are, perhaps, the most important factors that allow chickens to adjust to tropical environmental pressures.

Notably, among 10 identified genes in WCC selective pathways, we observed significantly higher Z (F_{ST}) and \log_2 ($\theta\pi$ ratio) values but lower Tajima's D values for the target gene *SLC33A1* compared with those in the adjacent genomic regions (Fig. 4B), indicating that a strong selective sweep occurred in this gene. *SLC33A1* encodes the acetyl-CoA (*Ac-CoA*) transporter (*ACATN*), which transports *Ac-CoA* from the cytoplasm to the golgi apparatus or the endoplasmic reticulum lumen that regulates protein acetylation [47]. *SLC33A1* promotes GD3 acetylation that contributes to mitochondrial damage repair in the glycosphingolipid biosynthesis-ganglio series pathway [48], which plays a crucial role in high temperature reactions [49]. Additionally, we observed that the gene for *thyroid stimulating hormone receptor* (*TSHR*), though not a candidate gene in the 5 selective pathways, had the strongest selective sweep (Fig. 4D). Also, it had markedly higher Z (F_{ST}) and \log_2 ($\theta\pi$ ratio) values but lower Tajima's D values in WCC compared with LDC in adjacent genomic regions. *TSHR* is the receptor for thyroid-stimulating hormone (TSH), the key regulator of thyroid function. *TSHR* has been reported to have a pivotal biological function in metabolic regulation during evolution [19,50].

We found within *SLC33A1*, which is highly conserved with particular significance [51], that specific alleles of three SNPs in WCC had lower mutation frequency that gradually became higher from GSC to LDC (Fig. 4C, supplementary table S16, Supplementary Material online), which was similar in pattern to each breed environmental location. Similar to changes in the allele frequencies of the three SNPs, the liver expression levels of *SLC33A1* were significantly higher in LDC than in WCC (Fig. 4K, Fig. 6). We also noticed that there were 2 SNPs within the most selective region of *TSHR* showed similar change law as the three SNPs of *SLC33A1* (Fig. 4E). Meanwhile, the interspecies NJ tree of *TSHR* sequences exposed that an evolutionarily conserved nucleotide has mutated which induced a change of coded amino acid (G to R) in WCC but remained variant in LDC and other breeds (supplementary Fig. S10, Supplementary Material online). Previously, this mutation has been identified as contributing to chicken evolution [50]. We further examined the mRNA expression levels of *TSHR* in livers

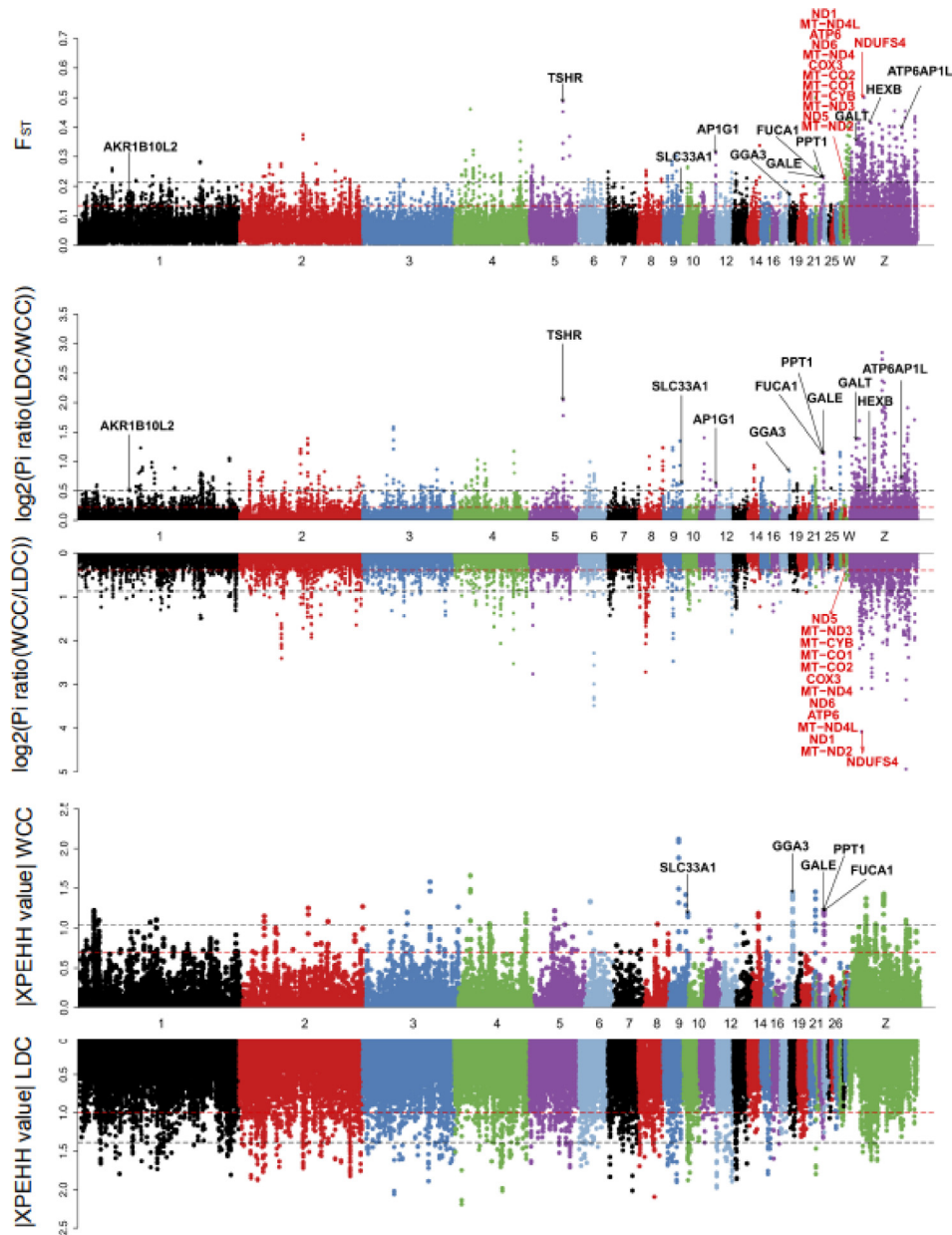


Fig. 3. Positive selection scans for adaptation to tropical and frigid environments. Chickens living in tropical environments (WCC) were compared with those living in frigid environments (LDC). The population genetic differentiation F_{ST} values, the nucleotide diversity $\theta\pi$ ratios ($\theta_{\pi-LDC}/\theta_{\pi-WCC}$) and $\theta\pi$ ratios ($\theta_{\pi-WCC}/\theta_{\pi-LDC}$) and the XP-EHH values are plotted in 40 kb genomic bins with a 20 kb step. The significance threshold of the selection signature was arbitrarily assigned to the top 5% percentile outliers for each individual test and is indicated with red horizontal dashed lines. The black horizontal dashed lines display the top 1% quantile. The black font genes represent the WCC selective signal, and the red font genes represent the LDC selective signal.

and its downstream genes *ADCY1* and *CREB3L3*. The data revealed that liver but expression levels of *ADCY1* and *CREB3L3* genes decreased as environmental temperature surrounding WCC to LDC decreased (Fig. 4K, Fig. 6). Although there was no change in RNA expression levels in *TSHR*, the expression changes of *ADCY1* and *CREB3L3* suggested the related signal pathway of *TSHR* might be activated in liver in response to temperature.

To further explore the SNPs identified above, we utilized SNPs data (chicken Affymetrix 600 K SNP chip) from 1,399 indigenous chickens from Tanzania (Eastern Africa) [52], and 1,439 indigenous chickens from Ghana (Western Africa) [53], and from 156 Brazilian chickens [54]. The SNPs allele frequencies of the selected 11 candidate genes (including *TSHR*) in these populations were calculated and compared with allele frequencies in the breeds of the current study. 3 mutations [2 in the *TSHR* gene (at position

40,827,385 and 40829460) and 1 in the *SLC33A1* gene (at position 23059382)] were found in the Affymetrix 600 k data, and their allele frequencies in these 2 large African populations (6 local ecotypes) supported their possible important roles in tropical environment adaption (Fig. 4F, 4I, 4J, supplementary table S16, Supplementary Material online).

To further examine whether allele frequency patterns of the 5 mutations in *SLC33A1* and *TSHR* were associated with tropical environments, 120 individuals originating from 4 Chinese breeds were genotyped by PCR (30 individuals per breed, supplementary fig. S11 and S12, supplementary table S17, Supplementary Material online). The collection test analyses were in support of a strong correlation between the frequencies of 3 variants and environment temperature (position 40827385, P -value = 1.05E-4; position 40829460, P -value = 1.11E-4, position 23059382, P -value = 0.03).

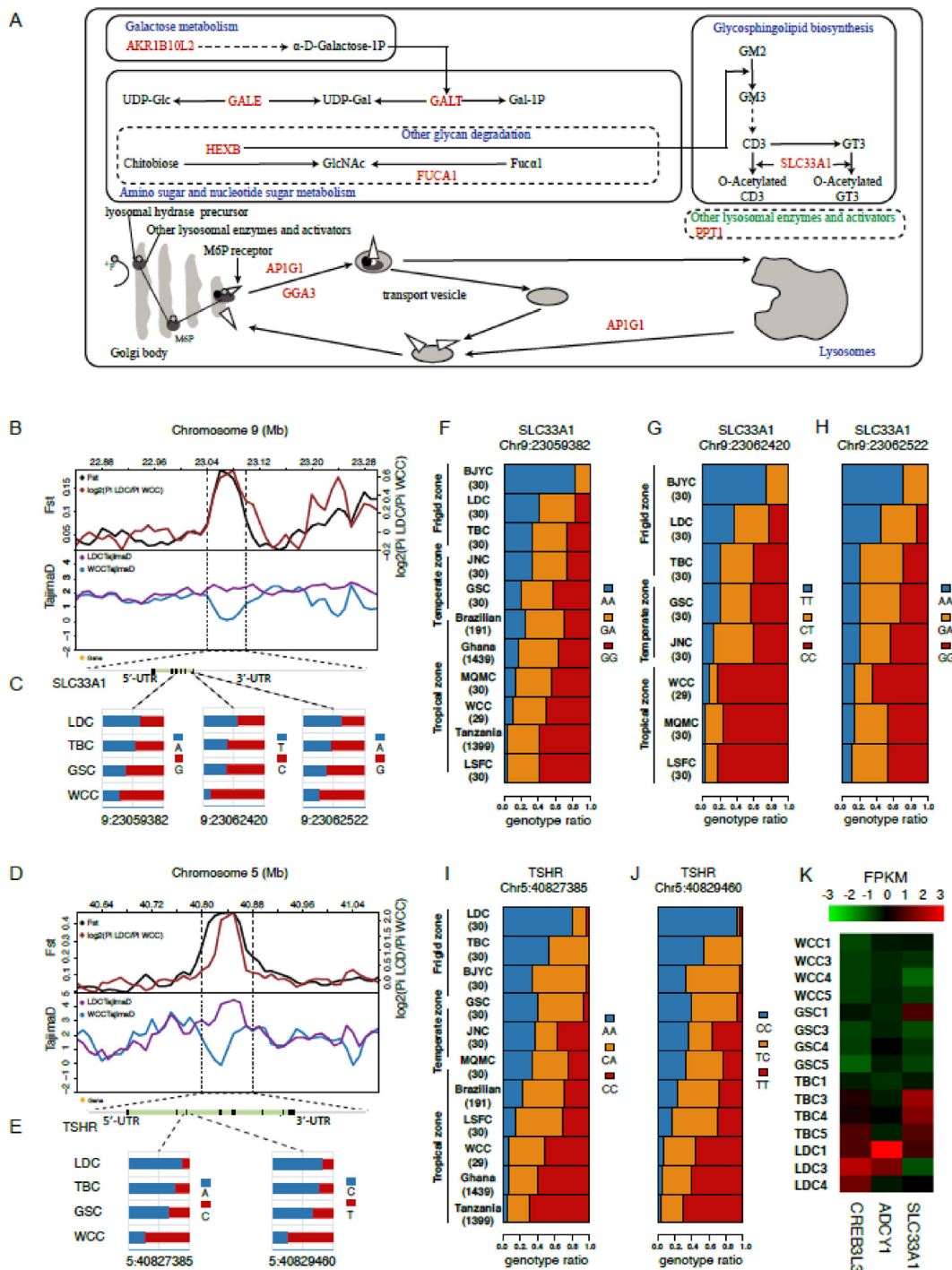


Fig. 4. Analysis of the signatures of positive selection in the genome of chicken breeds and the adaptations to tropical Environments. (A) Schematic mechanisms of signaling pathways for genetic adaptations to tropical environments in chicken. The names of the KEGG pathways are shown in blue. The candidate genes positively selected in the two methods of F_{ST} and θ_{π} ratio tests are shown in red. Dotted arrows indicate an indirect effect. (B) (D) $\log_2(\theta_{\pi}$ ratios), F_{ST} value and Tajima's D values around the *SLC33A1* locus, *TSHR* locus, respectively. The black and red line represent F_{ST} and $\log_2(\theta_{\pi}$ ratios) values, respectively. The blue and purple lines represent the WCC, LDC Tajima's D values, respectively. (C) (E) Allele frequencies of two and three SNPs within the *SLC33A1*, *TSHR* gene across the four breeds, respectively. The red alleles are consistent with reference genome, and the blue alleles represent mutant alleles. (F)(G)(H) Genotype frequencies of three SNPs within the *SLC33A1* gene were determined in our 119 individuals and chickens from Tanzania, Ghana, and Brazil using the Affymetrix 600 k SNP database, as well as in four Chinese breeds with PCR. For the abbreviations of the breeds, see supplementary table S16 and Supplementary Material online. (I)(J) Genotype frequencies of two SNPs within the *TSHR* gene were determined in our 119 individuals and chickens from Tanzania, Ghana, and Brazilian using the Affymetrix 600 k SNP database, as well as in four Chinese breeds with PCR. For the abbreviations of the breeds, see supplementary table S16 and Supplementary Material online. (E) Gene expression of *SLC33A1*, and the gene expression of the *TSHR* downstream genes, including *ADCY1* and *CREB3L3* in chicken liver of four breeds based on RNA-seq analysis. The FPKM (fragments per kilobase of transcript per million mapped reads) value is used to measure the expression level.

WCC living at > 25 °C showed the lowest allelic frequencies of the three mutations (~0.25), whereas LDC living in a low-temperature environment (<10 °C) showed the highest (~0.60–0.90). The other

chicken breeds presented corresponding frequencies (Fig. 4F–J, supplementary table S16, Supplementary Material online).

Adaptive mechanisms in frigid environments

For the purpose of understanding adaptive mechanisms in frigid environments, LDC, TBC, and GSC were compared to WCC, and the last 2 groups for comparison were set as the control group (supplementary Fig. S13, Supplementary Material online). Using the top 5% of both F_{ST} and $\theta\pi$ ratio cutoffs, we identified a 33.40 Mb of 835 selective sweep region encompassing 504 genes (Fig. 3, supplementary table S18, Supplementary Material online). Additionally, several highly enriched pathways were identified, including cardiac muscle contraction, oxidative phosphorylation, metabolic pathways, and a notch signaling pathway (supplementary table S21, Supplementary Material online). In the positive control group (TBC vs WCC) and the negative control group (GSC vs WCC), we discovered a 30.72 Mb of 768 selective sweep regions encompassing 407 genes with 5 highly enriched pathways (supplementary Fig. S14, supplementary table S19, Supplementary Material online) and a 44 Mb of 1100 selective sweep regions encompassing 564 genes with 7 highly enriched pathways (supplementary Fig. S15, supplementary table S20, Supplementary Material online). Interestingly, cardiac muscle contraction, oxidative phosphorylation and notch signaling pathways were selected pathways in both LDC and TBC but not in GSC, suggesting these three pathways' crucial role in adaptation to frigid environments. In total, 13 common genes in cardiac muscle contraction and oxidative phosphorylation pathways were found in LDC and TBC (Fig. 3, supplementary table S21, Supplementary Material online), which suggests their key status. Meanwhile, we found no overlapping genes in notch signaling pathways between LDC and TBC, indicating their different genetic mechanisms of frigid environment adaption. In addition, the folate biosynthesis pathway was selected in TBC, which has been identified as being related to plateau adaptability [55].

Alterations in the oxidative phosphorylation efficiency of mitochondria could play a crucial role in increased heat production of animals and plants exposed to cold [56,57], because this efficiency is necessary to maintain energy homeostasis. Along with increased oxidative phosphorylation, oxygen consumption increases, which modulates cardiac muscle contractility and adjusts cardiac output and pulmonary artery blood flow to increase the internal and external gas exchange rate [58]. In total, 13 genes notably gathered in the oxidative phosphorylation pathway as well as cardiac muscle contractility pathway are mainly oxidation–reduction enzyme genes in the mitochondrial genome, and they are considered important in adaptation to cold environments [59].

The target gene, *NADH dehydrogenase ubiquinone Fe-S protein 4* (*NDUFS4*), had the largest $Z(F_{ST})$ value among these 13 genes, with much higher $Z(F_{ST})$ and $\log_2(\theta\pi)$ values but lower Tajima's D values in LDC (Fig. 5A), suggesting a strong selective sweep. *NDUFS4* belongs to a group of karyogenes of mitochondrial complex I (CI), which is responsible for regulation of CI activity and mitochondrial respiration [60]. Within the *NDUFS4* genes, specific alleles of two SNPs (Z:16123795, Z:16134664) in the LDC as well as in the TBC populations owned notably higher frequencies than those observed in WCC. The other SNP (Z: 16145147) exhibited an opposite trend simultaneously (Fig. 5B).

Next, we sought to further investigate the potential functional consequences of SNPs from the gene *NDUFS4*. We studied whether they indicated any connection with frigid environments, stretching our analysis to other populations including 20 individuals from Poland, 92 individuals from Finland, 20 individuals from Israel and 37 individuals from Egypt [61,62]. We found the allele frequencies of three mutations of *NDUFS4* in Poland, Finland, Israel and Egypt population displayed similar variation tendency as the populations of WCC, GSC, TBC and LDC have suggested, based on their respective ambient temperatures. These results suggest strong correlation between the variants frequencies and environ-

mental temperature (Fig. 5C-E, supplementary table S22, Supplementary Material online).

Discussion

In this research, we sequenced the genomes of 119 chickens from four Chinese local breeds spanning a wide range of environmental temperatures, from tropical to frigid. A total of ~ 11.69 million SNPs (~2.21 million indels and SVs) were identified based on our vast genome data, which helped uncover the population structure of these Chinese local chickens. The location and time of chicken domestication in China has remained controversial even in recent years [63]. However, in line with previous work based on mitochondrial genomes of the chicken [9], our population genetic diversity analysis revealed some chickens from Southern China originated earlier than those from Northern China.

Sweeping the genome of the WCC breed for positive selection signatures to tropical environments, we uncovered a series of genes involved in a few major signaling pathways that were related to glycometabolism and energy metabolism and that played central roles in regulating cellular responses to tropical environments [46]. Within the top-candidate genes, *SLC33A1* is a key member of the endoplasmic reticulum (ER) acetylation machinery that belongs to the SLC family [64]. The SLC family facilitates transfer of various molecules, including sugars, nucleotides, and amino acids to maintain a constant internal environment [65]. Several SLC members (*SLC13A1*, *SLC5A11*) have been mined and described in connection with tropical environment adaption based on microarray analyses [66]. Additionally, some other members including *SLC7A7*, *SLC2A3*, *SLC16A1* and *SLC16A7* were proposed as having a high potential as heat stress biomarkers for different organisms [67,68]. Remarkably, *SLC33A1* is located in the glycosphingolipid biosynthesis-ganglio series pathway, one of the enriched pathways in our analysis, which regulates ganglioside GD3 and GT3, both are acidic glycosphingolipids that are important lipid and glycolipid mediators of the adaptive responses to various stressors [48], and it was also selected in Hainan pigs (Ding'an and Tunchang pigs) [69], which are exposed to warm environmental conditions; and, therefore, it is reasonable to assume that *SLC33A1* is important for adaption to tropical environments. The 3 SNPs within *SLC33A1* were found at lower mutation frequencies in chicken breeds from tropical zones than from temperate and frigid zones. This data was collected across a large quantity and different varieties of chickens, that were derived initially from Southeast Asia and that have a high adaptability to tropical environments [9]. Meanwhile, the expression levels of *SLC33A1* in the WCC breed were lower than in other breeds, because of its downregulation results in widespread cell death and induction of features characteristic of autophagy to maintain ER homeostasis [64].

It's also worth noting that *TSHR*, a critical gene with a strong selective sweep, has been described as a "domestication locus" in the chicken [19,70]. Functional consequences of mutations in the *TSHR* gene remain to be explored, though previous studies have confirmed a pivotal mediating role for *TSHR* in reducing aggression to conspecifics, decreasing ascites, controlling reproductive machinery, and influencing broodiness [19,70,71]. Interestingly, in our study *TSHR* appears to be involved in high temperature adaptation, one of the prevailing environmental stressors for which natural selection footprints are often associated [72]. The *TSHR* SNP variant exhibited a glycine-to-arginine substitution that was unique to the WCC breed, and it was almost completely conserved among the other chicken breeds (supplementary Fig. S10, Supplementary Material online). Currently, this identified missense substitution has not revealed any indication that it is a deleterious mutation. It is interesting to note that lower mutation frequencies

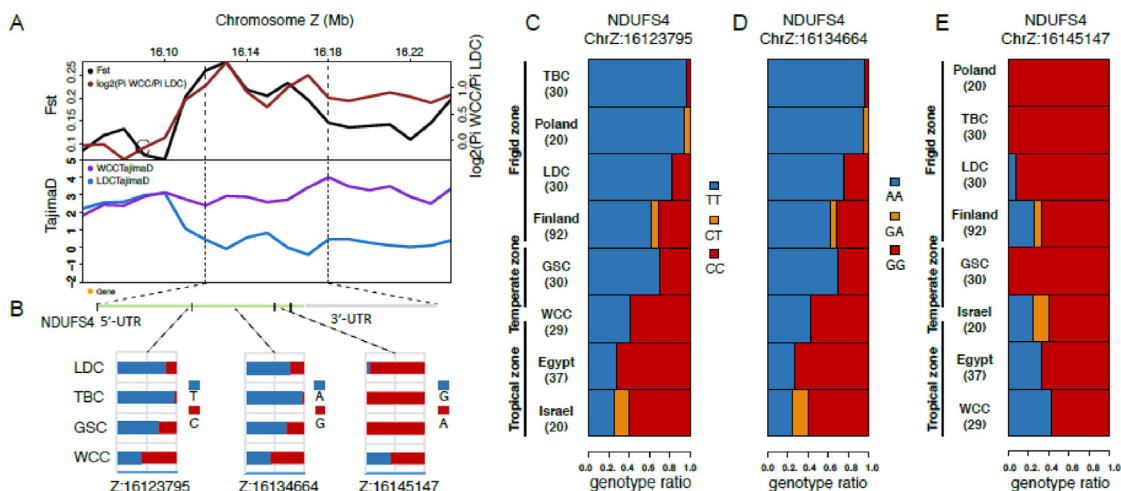


Fig. 5. Analysis of the signatures of positive selection in the genome of chicken breeds and adaptations to frigid environments. (A) $\log_2(\theta_{PI} \text{ ratios})$, F_{ST} value and Tajima's D values around the *NDUFS4* locus. The black and red line represent F_{ST} and $\log_2(\theta_{PI} \text{ ratios})$ values, respectively. The blue and purple lines represent the WCC, LDC Tajima's D values, respectively. (B) Allele frequencies of three SNPs within the *NDUFS4* gene across the four breeds, respectively. The red alleles are consistent with reference genome, and the blue alleles represent mutant alleles. (C)(D)(E) Genotype frequencies of three SNPs within the *NDUFS4* gene were identified in our 119 individuals and in chickens from Poland, Finland, Israel and Egypt. For the abbreviations of the breeds, see supplementary table S22 and Supplementary Material online.

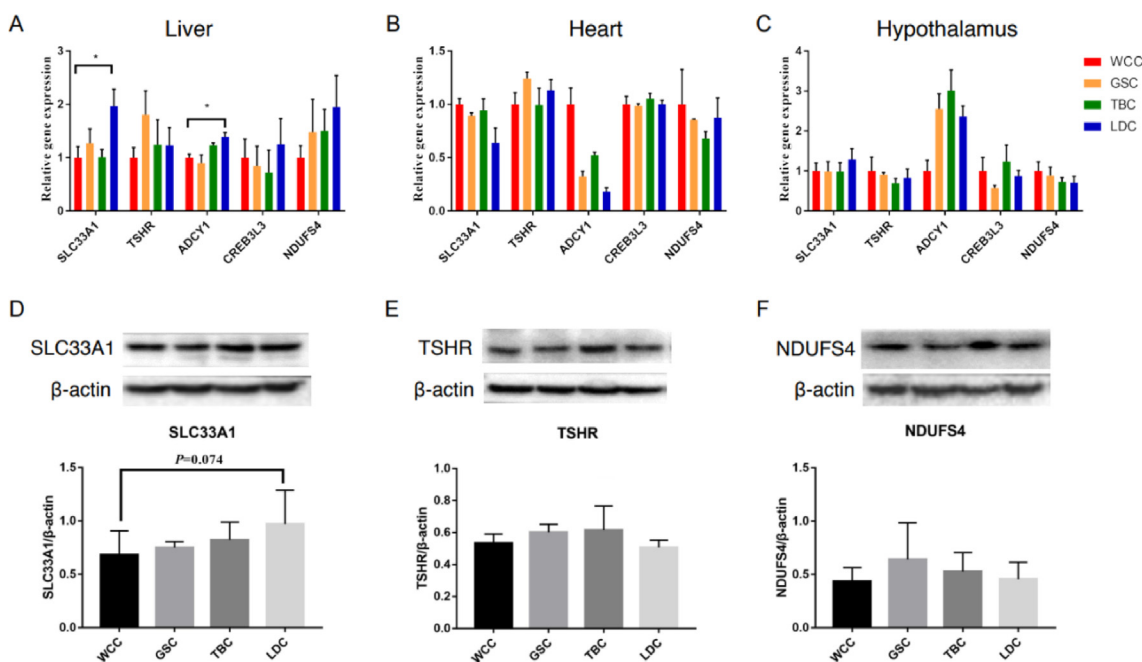


Fig. 6. Validation assays of the selective genes for adaptation to tropical and frigid environments. (A)(B)(C) Temporal relative expression of five genes in liver, heart, hypothalamus across the four breeds measured by qRT-PCR, respectively. (D)(E)(F) Western-blotting analysis of *SLC33A1*, *TSHR*, *NDUFS4*.

of the other two SNPs (5: 40827385, 5: 40829460) within *TSHR* were discovered in chicken breeds from tropical zones, and these 2 *TSHR* SNPs exhibited a similar variation tendency to that of the 3 SNP mutations of *SLC33A1*. We speculate that the *TSHR* allele exhibits an advantage of increasing metabolic activity and growth. We intend to follow up on the discovery of the *TSHR* sweep with functional receptor studies.

Genetic variation in the genome of the LDC breed, which is characterized by adaptations to frigid environments, particularly evolved genes associated with mitochondria oxidative phosphorylation (OXPHOS) in the heart that will ultimately generate heat to make an organism more resistant to low temperatures [73]. Several selected genes, including *ND2*, *ND3*, *ND4*, *ND5*, *MT-CYB* and *ATP6*, are known to be important for adaptation to frigid environments in

many organisms [74–76]. Variations in these genes are speculated to influence the mitochondrial coupling efficiency and coordination of the ATP synthesis efficiency ratio to thermogenesis efficiency, which assists in climate adaptation [77]. In addition, a NADH-ubiquinone oxidoreductase gene, *NDUFS4*, was notably found to exhibit a strong selective sweep in our results. A previous study revealed mutations in *NDUFS4* can enhance cold, salt, and osmotic stress tolerance in Arabidopsis [78]. Moreover, *NDUFS4* can catalyze electron transport coupled with proton translocation across the inner mitochondrial membrane, thus providing the driving force for ATP synthesis and antioxidant defense maintenance [79], which are strongly related to the cold stress response [80]. In our studies, higher frequencies of 3 SNP variants in the conserved region of *NDUFS4* were observed in chickens from frigid environments (LDC

and TBC), as well as chickens from Poland and Finland, but not in the more tropical WCC breed, and chickens from Israel and Egypt. These results suggest the value of these detailed advantageous alleles for adaptation of chickens to the frigid environments.

Conclusion

In conclusion, the present study provided a whole-genome sequence analysis of Chinese native chicken breeds. The analysis included 119 individuals of 4 breeds from tropical environments to frigid environments. Patterns of genetic diversity revealed differentiation as well as close relationships among these breeds following domestication. Genomic regions exhibited a signature of positive selection revealing unique difference due to selection pressure of the environment. The more tropical WCC revealed a stronger selection toward a variety of novel genes located in 5 pathways that mainly participate in glycometabolism and energy metabolism, which are potentially involved in adaptation to tropical environments. While the LDC from frigid regions exhibited more selection pressure toward cardiac muscle contraction, oxidative phosphorylation and notch signaling pathways related to energy homeostasis in response to frigid environments. The *SLC33A1* gene displayed a stronger signature for positive selection in WCC, with lower mutation frequency in 3 selected SNPs and lower expression in the liver than in the other breeds. Another gene, *TSHR*, exhibited the same nonsynonymous mutation as has been found in previous studies, and it also exhibited 2 other SNPs with lower mutation frequency in WCC. 6 SNPs were confirmed in 1395 Tanzania and 1439 Ghana local African chickens, 156 chickens from Brazil, and 120 individuals of 4 Chinese breeds; and a lower mutation frequency was obtained in these chickens from tropical environments. Additionally, the *NDUFS4* gene presented a higher frequency of SNPs in breeds from LDC and TBC as well as other chickens from the colder environments of Poland and Finland. Overall, these results indicate that the genetic differences seen in these very distinct populations likely contribute to their tolerance of and survival in extreme environments. Our genomic data formed in this research will be of value for genetic breeding of chickens that will face extreme climate change.

Compliance with ethics requirements

All Institutional and National Guidelines for the care and use of animals were followed.

Data availability

The accession numbers for genome re-sequencing data and transcriptome data reported in this paper are NCBI SRA: PRJNA800119. Other data that supported the discoveries could be obtained from the corresponding authors.

CRediT authorship contribution statement

Shourong Shi: Conceptualization, Methodology, Visualization, Writing – review & editing. **Dan Shao:** Visualization, Writing – review & editing. **Lingyun Yang:** Software, Data curation. **Qiqi Liang:** Software, Data curation. **Wei Han:** Investigation. **Qian Xue:** Software, Data curation. **Liang Qu:** Investigation. **Li Leng:** Investigation. **Yishu Li:** Investigation. **Xiaogang Zhao:** Investigation. **Ping Dong:** Investigation. **Muhammed Walugembe:** Data curation. **Boniface B. Kayang:** Data curation. **Amandus P. Muhairwa:** Data curation. **Huaijun Zhou:** Methodology, Writing – review & editing. **Haibing Tong:** Conceptualization, Methodology, Visualization, Writing – review & editing.

Declaration of Competing Interest

The authors declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper.

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Appendix A. Supplementary material

Supplementary data to this article can be found online at <https://doi.org/10.1016/j.jare.2022.07.005>.

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