

In silico approaches to discover the functional impact of non-synonymous single nucleotide polymorphisms in selective sweep regions of the Landrace genome

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Objective: The aim of this study was to discover the functional impact of non-synonymous single nucleotide polymorphisms (nsSNPs) that were found in selective sweep regions of the Landrace genome

Methods: Whole-genome re-sequencing data were obtained from 40 pigs, including 14 Landrace, 16 Yorkshire, and 10 wild boars, which were generated with the Illumina HiSeq 2000 platform. The nsSNPs in the selective sweep regions of the Landrace genome were identified, and the impacts of these variations on protein function were predicted to reveal their potential association with traits of the Landrace breed, such as reproductive capacity.

Results: Total of 53,998 nsSNPs in the mapped regions of pigs were identified, and among them, 345 nsSNPs were found in the selective sweep regions of the Landrace genome which were reported previously. The genes featuring these nsSNPs fell into various functional categories, such as reproductive capacity or growth and development during the perinatal period. The impacts of amino acid sequence changes by nsSNPs on protein function were predicted using two in silico SNP prediction algorithms, i.e., sorting intolerant from tolerant and polymorphism phenotyping v2, to reveal their potential roles in biological processes that might be associated with the reproductive capacity of the Landrace breed.

Conclusion: The findings elucidated the domestication history of the Landrace breed and illustrated how Landrace domestication led to patterns of genetic variation related to superior reproductive capacity. Our novel findings will help understand the process of Landrace domestication at the genome level and provide SNPs that are informative for breeding.

Keywords: Landrace; Next-generation Sequencing; Non-synonymous Single Nucleotide Polymorphism; Reproductive Capacity; Selective Sweep

INTRODUCTION

The recently developed high-throughput and cost-effective genotyping techniques allow the thorough exploration of genetic variation in domestic animals. In particular, whole-genome sequencing is a powerful approach for detecting massive amounts of single nucleotide polymorphisms (SNPs) in genome-wide sequence data. One of the strategies for studying genetic variation is to detect the selective sweep signatures based on patterns of linkage disequilibrium (LD) [1], which was proposed by Smith and Haigh [2], and other researchers have expanded and applied it [3-6]. Wang et al [7] performed a relative extended haplotype homozygosity (REHH) test to detect selective sweep regions of the Landrace genome using genotyping by genome sequencing. The genetic signature for selection of body size investigated by estimating the XP-EHH statistic in the Yucatan miniature pig [8]. Whole-genome re-sequencing of Jeju black pig (JBP) and Korean native pigs (which live on the Korean peninsula) were

performed to identify signatures of positive selection in JBP, the true and pure Korean native pigs [9]. Studies of selective sweeps in pigs have revealed strong selection signatures associated with genes underlying economic traits such as the body length, disease resistance, pork yield, muscle development, and fertility [10,11].

Diverse types of variants, e.g. copy number variations, insertion/deletion (InDel) and structural variations, have been identified in the selective sweep regions of the Landrace genome [7]. Unlike many SNPs are phenotypically neutral, nonsynonymous SNPs (nsSNPs) that are located in protein-coding regions and lead to amino acid substitutions in the corresponding protein product might have functional impacts and play a role in biological processes through altering the protein structure, stability, or function, these variations are often strongly associated with several phenotypes [12]. In the case of pigs, previous studies reported the different polymorphic patterns of nsSNPs in the Toll-like receptor genes between European wild boars and domestic pigs [13].

In this study, we aimed to identify nsSNPs in the selective sweep regions of the Landrace genome that might be related to superior reproductive capacity or growth and development during the perinatal period, and gene networks that were enriched in Landrace genome. Finally, impact of amino acid changes by nsSNPs on protein function was also investigated using *in silico* bioinformatic tools.

MATERIALS AND METHODS

Sample preparation and whole-genome re-sequencing In this study, a whole-genome sequence data set consisting of 14 Landrace (Danish), 16 Yorkshire (Large White) pigs, and 10 wild boars, were obtained from the NCBI Sequence Read Archive database (SRP047260). FastQC software [14] were used to perform a quality check on raw sequence data. Using Trimmomatic-0.32 [15], potential adapter sequences were removed before sequence alignment. Paired-end sequence reads were mapped to the pig reference genome (Sscrofa 10.2.75) from the Ensembl database using Bowtie2 [16] with the default settings. For downstream processing and variant calling, following software packages were used: Picard tools (http:// broadinstitute.github.io/picard/), SAMtools [17], and Genome Analysis Toolkit (GATK) [18]. "CreateSequenceDictionary" and "MarkDuplicates" Picard command-line tools were used to read reference FASTA sequences for writing bam files with only a sequence dictionary and to filter potential polymerase chain reaction duplicates, respectively. Using SAMtools, index files were created for the reference and bam files. Local realignment of sequence reads was performed to correct misalignment due to the presence of small insertions and deletions using GATK "Realigner-TargetCreator" and "IndelRealigner" arguments. In addition, base quality score recalibration was performed to obtain accurate quality scores and to correct the variation in quality with machine cycle and sequence context. For calling variants, GATK "UnifiedGenotyper" and "Select-Variants" arguments were used with the following filtering criteria. All variants with i) a Phred-scaled quality score of less than 30; ii) read depth less than 5; iii) MQ0 (total count across all samples of mapping quality zero reads) >4; or iv) a Phred-scaled p-value using Fisher's exact test of more than 200 were filtered out to reduce false-positive calls due to strand bias. "vcf-merge" tools of VCFtools [19] were used to merge all of the variants calling format files for the 40 samples. Additionally, tri-allelic SNPs were excluded, and all filtered SNPs on autosomes (a total of 26,240,429 SNPs) were annotated using an SNP annotation tool, SnpEff version 4.1a and the Ensemble Sus scrofa gene set version 75 (Sscrofa10.2.75). 53,998 nsSNPs (missense variants) were identified on autosomes from 40 sets of pig whole-genome data (Figure 1). Then, certain SNPs due to poor genotyping quality were removed; 4,174 SNPs were excluded based on Hardy-Weinberg equilibrium testing ($p \le 0.000001$). In addition, a total of 19,002 SNPs with a minor allele frequency of <0.05 were excluded. After genomic data quality control, there were 30,822 SNPs for downstream analysis.

Population structure analysis

Population structure analysis was performed to infer the population structure of the 40 pigs with whole-genome sequence data. The program STRUCTURE (https://web.stanford.edu/ group/pritchardlab/structure.html) was used to evaluate the extent of substructure among the 40 individuals belonging to three pig breeds. Bayesian clustering analysis implemented in STRUCTURE (version 2.3.4) was used to estimate the population structure using 30,822 nsSNPs from the whole-genome sequencing data of the 40 pigs [20]. An initial burn-in of 10,000 iterations were followed by 10,000 iterations for parameter estimation was sufficient to ensure the convergence of parameter estimates. To estimate the number of populations (the K parameter of STRUCTURE), the dataset was analyzed by allowing for the values of K = 3 (Figure 2).

Identify nsSNPs in Landrace selective sweep regions

A previous study identified 269 selective sweep regions of the Landrace genome using the REHH test (p-value≤0.01), which was used to detect the recent positive selection signatures by evaluating how LD decays across the genome 7. A total of 261 of 269 selective sweep regions of the Landrace genome were on autosomes, and 345 nsSNPs belonged to 55 Landrace selective sweep regions were identified (Figure 3). Overall, 345 nsSNPs in 55 selective sweep regions of the Landrace genome belonged to 90 genes, and gene function 64 of total 90 genes were discovered. Gene ontology (GO) network analysis was performed using ClueGO [21] to infer the biological mean-

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Figure 1. Functional classification of total single nucleotide polymorphisms (SNPs) from 40 pig whole-genome sequences (16 Yorkshire, 14 Landrace, and 10 wild boar). After SNP calling, all filtered SNPs (a total of 26,240,429 SNPs) were annotated using an SNP annotation tool, SnpEff version 4.1a (reference), and the Ensembl *Sus scrofa* gene set version 75 (Sscrofa10.2.75). Through SnpEff, we divided all SNPs into 31 functional classes containing non-synonymous SNPs (missense variants), as shown in this figure. The dotted line box in this figure indicates non-synonymous SNPs.

ing of the genes related to nsSNPs in Landrace selective sweep regions.

Predicting damaging amino acid substitutions of non-synonymous SNPs specific to the Landrace breed

In this study, the functional effects of nsSNPs were predicted using the following in silico algorithms: sorting intolerant from tolerant (SIFT) [22] and polymorphism phenotyping v2 (Polyphen-2) [23]. Total 345 nsSNPs in 55 selective sweep regions of the Landrace genome were analyzed using SIFT. NsSNPs with less than 0.05 of SIFT score, which was regarded as deleterious, were used for PolyPhen-2 ver. 2.2.2 (http://genetics. bwh.harvard.edu/pph2/) analysis to predict the influence of an amino acid change on the structure and function of a protein by using specific empirical rules [23]. From the results of Polyphen-2 analysis, nsSNPs were classified into probably damaging, possibly damaging, and benign based on their scores (ranging from 0 to 1); if Polyphen-2 score for nsSNPs was more than 0.95, nsSNPs were considered to be "probably damaging", while for values between 0.5 and 0.95, they were regarded as "possibly damaging". The scores below 0.5 were classified as "benign". In this study, probably damaging and possibly damaging SNPs were judged as to have strong effects on protein function.

If the SIFT score of each SNP was less than 0.05, the SNP was regarded as being deleterious, which could strongly affect protein function. Additionally, we performed PolyPhen-2 (version 2.2.2) analysis to predict the influence of an amino acid change on the structure and function of a protein by using specific empirical rules [23]. Amino acid sequences corre-





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Figure 3. Genotypes of 345 non-synonymous single nucleotide polymorphisms (SNPs) in Landrace selective sweep regions. The genotype patterns of 345 non-synonymous SNPs in the selective sweep regions of the Landrace genome are represented by a heat map. The colors of the boxes represent the genotypes of each of the 40 individuals from the whole-genome sequencing data. Dark blue indicates that the genotypes of both the alleles were the same as that of the minor allele. Blue boxes indicate that one of the two alleles was the same as the minor allele and the other was the same as the major allele. Sky blue means that the genotypes of both alleles were the same as that of the figure shows a list of each SNP name, which consists of the chromosome, position, and minor allele type. The gray box at the bottom of the figure indicates the three breeds.

sponding to nsSNPs of interest from the Ensembl database were obtained to perform PolyPhen-2 analysis.

RESULTS

DNA sequencing, data preprocessing, and genetic variant calling

A total of 26,240,429 SNPs were extracted on autosomes from the whole-genome sequences of the 40 pigs, including 14 Landrace individuals, and annotated all extracted SNPs using SnpEff version 4.1a (http://snpeff.sourceforge.net/SnpSift.html) [24]. Through this SNP annotation, all SNPs were divided into 31 functional classes, including nsSNPs (Figure 1). Most of the SNPs were located in intergenic or intronic regions; finally, we identified 53,998 nsSNPs (0.205% of the total SNPs). After quality control for all of the nsSNPs, there were 30,822 nsSNPs. Population structure analysis using the genotypic information on these SNPs provided the genetic relationship among breeds. The results from analyzing the population structure clearly distinguished Landrace, Yorkshire, and wild boar (Figure 2).

nsSNPs in Landrace selective sweep regions

A total of 269 selective sweep regions were obtained from a previous study on the Landrace breed to identify nsSNPs related to selective sweeps [7], and a total of 345 nsSNPs were identified from 55 Landrace selective sweep regions (Figure 3) by re-analyzing the data of previous study resequencing data of Landrace and Yorkshire [7]. Information of 345 nsSNPs in the selective sweep regions of the Landrace genome belonged to 90 genes were shown in Table 1. The average number of nsSNPs per gene was 3.83, and the gene length was not correlated to the number of nsSNPs (Figure 4). The deleted in malignant brain tumors 1 (DMBT1) gene consisted of 18 exons harboring 26 nsSNPs that were evenly distributed; this gene had the highest number of nsSNPs among the 90 genes. Moreover, there were considerable frequency differences between Landrace and other breeds (Yorkshire and wild boar) in nsSNPs of the DMBT1 gene (Figure 5). This suggests that

Table	1.	Gene list	containing	non-synonymous	SNPs in	Landrace	selective	sweep regions
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Gene name	CHR	Gene sart	Gene end	# ns SNP	Selective sweep region	Gene name	CHR	Gene sart	Gene end	# ns SNP	Selective sweep region
PLG	1	8,739,981	8,787,582	8	1:8670943-8797806	ENSSSCG00000015184	9	56,925,449	56,927,199	4	9:56869539-57122277
MELK	1	265,175,024	265,288,283	1	1:265063188-265212930	ENSSSCG00000026119	9	56,962,203	56,963,135	7	
ZFPL1	2	6,231,271	6,235,566	1	2:6227731-6239068	ENSSSCG00000015182	9	56,971,208	56,972,140	5	
ENSSSCG00000021162	2	15,576,680	15,577,609	3	2:15569156-15593980	ENSSSCG00000028463	9	56,980,334	56,981,572	5	
FAM180B	2	16,204,579	16,206,256	3	2:16111708-16299440	ENSSSCG00000024117	9	57,283,042	57,284,501	5	9:57230656-57379772
ENSSSCG00000025219	2	62,507,452	62,508,408	1	2:62355986-62756249	ENSSSCG00000024455	9	57,293,941	57,296,806	1	
ENSSSCG00000013821	2	62,624,616	62,625,548	14		DMTF1	9	102,893,256	102,929,921	1	9:102847568-103896296
ENSSSCG00000013822	2	62,644,870	62,645,796	14		DENND1B	10	25,096,498	25,193,569	1	10:25139986-25249094
ENSSSCG00000013819	2	62,669,703	62,670,662	8		ENSSSCG00000010907	10	26,249,079	26,284,300	17	10:26197521-26710943
MCOLN1	2	72,056,664	72,151,713	1	2:72143419-72172550	PTPRC	10	26,308,759	26,332,284	2	10:26197521-26710943
ENSSSCG00000014078	2	85,731,838	85,732,242	4	2:85467258-86506548	KIAA1462	10	45,386,450	45,428,443	4	10:45403837-45436342
ANKRD31	2	85,774,886	85,807,199	2		GJD4	10	63,677,681	63,683,060	2	10:63669866-63725092
ANKDD1B	2	86,257,325	86,321,705	3		ENSSSCG00000021829	11	11,141,413	11,236,840	1	11:10400737-11376721
SDK1	3	3,634,288	3,824,252	3	3:3730382-3773007	ENSSSCG00000020699	11	11,355,261	11,378,042	1	
PLA2G6	5	6,996,414	7,059,756	1	5:6988526-7058468	CCDC168	11	78,361,372	78,368,847	22	11:78318648-78678168
BIN2	5	17,315,117	17,339,457	1	5:17248525-17487183	DNAI2	12	6,779,152	6,799,278	3	12:6771152-6805468
TAC3	5	24,048,553	24,056,427	3	5:23288996-24074802	MARCH10	12	15,897,681	15,944,341	10	12:15890650-15938045
ZBTB39	5	24,066,660	24,068,784	4		MAPT	12	17,123,471	17,172,747	2	12:16937097-17191735
NCAPD2	5	66,432,584	66,443,844	1	5:66396846-66725591	CCL23	12	41,160,877	41,165,234	3	12:41158920-41165901
VAMP1	5	66,646,135	66,647,743	1		CCL1	12	42,467,618	42,471,014	3	12:42468535-42621081
TAPBPL	5	66,647,211	66,658,624	2		ENSSSCG00000017834	12	50,542,085	50,552,985	1	12:50535159-50581774
DMBT1	6	43,728,925	43,753,137	26	6:43719388-43757067	SHPK	12	51,572,871	51,592,551	1	12:51579885-51586595
ENSSSCG00000027618	6	119,199,612	119,199,920	3	6:119198939-119344591	SPNS3	12	52,389,071	52,445,090	1	12:52401285-52444137
MCOLN2	6	119,212,826	119,273,364	1		CCDC66	13	42,284,163	42,341,496	2	13:41196871-42465605
PCNX1	7	100.745.867	100.862.081	2	7:100703442-100775415	NOC4L	14	24,724,492	24,730.021	2	14:24592939-24779049
PLD4	7	131,340,863	131,347,987	3	7:131291714-131388688	DDX51	14	24,730,045	24,732,878	2	
ENSSSCG0000002551	7	131,356,311	131,359,461	5		EP400	14	24,748,336	24,847,567	2	
ATP8A1	8	35,180,992	35,309,867	2	8:34998191-35275833	ENSSSCG00000010013	14	50,652,381	50,652,947	1	14:50647172-50719083
ENSSSCG00000027999	9	2,277,256	2,278,264	7	9:2223331-2577505	OSBP2	14	50,669,019	50,849,290	2	
OVCH2	9	2.307.953	2.321.197	10		KIF20B	14	110.499.118	110.581.337	1	14:110280822-110542445
ENSSSCG00000025898	9	2.361.209	2.362.147	5		FGFR1IIIC	15	55.215.592	55,269,381	1	15:55142754-55608192
ENSSSCG00000023477	9	2.370.889	2.371.830	12		LETM2	15	55.274.276	55,294,333	1	
ENSSSCG00000029634	9	2.455.370	2.528.783	1		WHSC1L1	15	55.338.007	55,406,429	2	
TRIM3	9	3,923,986	3,940,046	1	9:3927497-3978728	DDHD2	15	55,414,565	55,455,195	1	
HPX	9	3,946,381	3.955.253	4		ASH2L	15	55.512.104	55,552,504	1	
SMPD1	9	3.961.589	3,964,504	1		ENSSSCG00000029683	15	128,593,493	128,594,377	6	15:128498493-128627886
MOGAT2	9	11.119.062	11.132.962	11	9:11120076-11136889	CWC27	16	46.572.512	46.875.541	2	16:46472193-46771773
THAP12	9	11.652.415	11.669.844	2	9:11449284-11760977	CD93	17	34,381,626	34,384,902	2	17:34206246-34400408
GAB2	9	13,936,307	14,135,685	- 1	9:13934282-14030509	GZF1	17	34.441.517	34.447.221	3	17:34421087-34505222
FI MOD1	9	40,189,956	40.282.814	1	9:40189621-40286365	NAPB	17	34,450,368	34.485.152	1	
ATM	9	40.925.895	40.945.439	3	9:40793693-41170478	CSTL1	17	34,492,910	34,496,585	2	
KDELC2	- 9	41.043.564	41.065.077	7		CST7	17	34.906.655	34,915,135	1	17:34901568-34908632
EXPH5	9	41 073 546	41 217 329	12		DEER119	17	39 921 302	39 931 655	2	17:39862221-40018288
ENSSSCG00000023913	9	41,145,017	41.152.176	3		DEFB116	17	39.996.662	39,999,076	1	
ARHGAP20	9	43,174,648	43,222,583	1	9:43134418-43291918	ENSSSCG0000007337	17	46,357,154	46,401,936	2	17:46275105:46424519

SNPs, single nucleotide polymorphisms; nsSNPs, non-synonymous SNPs. We show the information of genes containing non-synonymous SNPs. In this table, the fifth column indicates the number of non-synonymous SNPs in each gene and the seventh column presents information on the selective sweep regions of the Landrace genome and selective sweep name consisting of thromosome, start position, and end position.

DMBT1 is significantly affected by many nsSNPs in Landrace breed establishment. Previous studies strongly suggested an important role of DMBT1 in the process of fertilization in pigs; it was shown to be secreted in the oviduct and involved in the mechanism of fertilization in porcine species [25,26]. In particular, Ambruosi et al [25] reported that oviduct fluid containing DMBT1 protein was strongly related to the preparation of gametes for fertilization, fertilization itself, and subsequent embryonic development. Therefore, we assumed that nsSNPs of DMBT1 of Landrace might correlate with the fertilization capacity that was acquired during artificial selection, making the reproductive capacity of Landrace pigs superior to that of other breeds [27].

Among 90 genes, the functions of 64 genes were predicted, and we performed GO network analysis of these 64 genes using ClueGO [21] to draw inferences on the biological effects of nsSNPs in Landrace selective sweep regions. The information on these networks is shown in Figure 6 and Table 2. The GO

network analysis revealed that 19 of the total of 64 genes were associated with five major GO terms, and these major terms were closely related to the reproductive capacity or growth and development of the Landrace breed during the perinatal period. In the GO network, seven genes (C-C motif chemokine ligand 1 [CCL1], CCL23, hemopexin, mucolipin 1, leucine zipper and EF-hand containing transmembrane protein 2, phospholipase A2 group VI [PLA2G6], and protein tyrosine phosphatase, receptor type, C [PTPRC]) were related to cellular metal ion homeostasis in seven major GO terms, and this cluster was the largest in this network. Moreover, these terms were similar to the GO results of a positively selected region identified in Wang's study of Landrace selective sweeps [7]. Metal ions are one major group of mineral; since components of follicular fluid such as Ca, Cu, and Fe significantly increase as the follicles increase in size, some minerals appear to play an important role in pig reproduction [28]. Five genes (ATPase phospholipid transporting 8A1 [ATP8A1], CCL1,



Figure 4. Correlation between length and number of single nucleotide polymorphisms (SNPs) in genes related to non-synonymous SNPs (nsSNPs) in Landrace selective sweep regions.

kinesin family member 20B, plasminogen, and *PTPRC*) were shown to be involved in the positive regulation of locomotion, and its network consisted of four GO terms (positive regulation of locomotion, positive regulation of cellular component movement, positive regulation of cell motility, and positive regulation of cell migration). This cellular movement is a central process in the development and maintenance of multicellular

DMBT1

organisms. In addition, tissue formation during embryonic development requires the orchestrated movement of cells in a particular direction. It is reasonable to assume that several genes of these four significant GO terms in the selective sweep regions of the Landrace genome might be related to the superior growth and development of Landrace during the perinatal period. Ten genes (*ATP8A1*, bridging integrator 2, CD93 mole-



Figure 5. Frequency difference of non-synonymous single nucleotide polymorphisms (nsSNPs) in deleted in malignant brain tumors 1 genes between Landrace and other breeds (Yorkshire and wild boar).



Figure 6. Gene ontology (GO) network analysis of genes related to non-synonymous single nucleotide polymorphisms (SNPs) in Landrace selective sweep regions. Significant results of GO analysis using genes related to non-synonymous SNPs in the selective sweep regions of the Landrace genome with our criteria in ClueGO packages of Cytoscape (number of genes = 4, sharing group percentage = 40.0). These results are largely divided into eight clusters as follows.

cule [*CD93*], exophilin 5, GRB2 associated binding protein 2, n-ethylmaleimide-sensitive factor attachment protein, beta, *PLA2G6*, *PTPRC*, and vesicle associated membrane protein 1 [*VAMP1*]) were associated with exocytosis, and five genes (*ATP8A1*, *CD93*, *DMBT1*, *PTPRC*, and *VAMP1*) were classified under the secretory granule membrane term in the GO network. The acrosome contains a single secretory granule and is located in the head of mammalian sperm; secretion from this granule is an absolute requirement for fertilization [29]. Acrosome exocytosis is a synchronized and tightly regulated all-or-nothing process, which provides a unique model for studying the multiple steps of the membrane fusion cascade [29]. Therefore, we assumed that these genes containing nsSNPs in the selective sweep region, which are related to exocytosis and the secretory granule membrane, might have been influenced by artificial selection, considering the distinctive reproductive capacity of the Landrace breed [27].

GO ID	GO Term	Term p-value	Group p-value	#Genes	Associated genes found
GO:0002274	Myeloid leukocyte activation	0.005	0.005	7	ATP8A1, BIN2, CD93, GAB2, MAPT, PTPRC, SHPK
GO:0006887	Exocytosis	0.001	0.001	10	ATP8A1, BIN2, CD93, EXPH5, GAB2, NAPB, PLA2G6, PLG, PTPRC, VAMP1
GO:0030667	Secretory granule membrane	0.003	0.003	5	ATP8A1, CD93, DMBT1, PTPRC, VAMP1
GO:0040017	Positive regulation of locomotion	0.016	0.017	5	ATP8A1, CCL1, KIF20B, PLG, PTPRC
GO:0051272	Positive regulation of cellular component movement	0.013		5	ATP8A1, CCL1, KIF20B, PLG, PTPRC
GO:2000147	Positive regulation of cell motility	0.012		5	ATP8A1, CCL1, KIF20B, PLG, PTPRC
GO:0030335	Positive regulation of cell migration	0.010		5	ATP8A1, CCL1, KIF20B, PLG, PTPRC
GO:0006873	Cellular ion homeostasis	0.003	0.006	7	CCL1, CCL23, HPX, LETM2, MCOLN1, PLA2G6, PTPRC
GO:0055080	Cation homeostasis	0.005		7	CCL1, CCL23, HPX, LETM2, MCOLN1, PLA2G6, PTPRC
GO:0030003	Cellular cation homeostasis	0.003		7	CCL1, CCL23, HPX, LETM2, MCOLN1, PLA2G6, PTPRC
GO:0055065	Metal ion homeostasis	0.003		7	CCL1, CCL23, HPX, LETM2, MCOLN1, PLA2G6, PTPRC
GO:0072507	Divalent inorganic cation homeostasis	0.015		5	CCL1, CCL23, MCOLN1, PLA2G6, PTPRC
GO:0006875	Cellular metal ion homeostasis	0.001		7	CCL1, CCL23, HPX, LETM2, MCOLN1, PLA2G6, PTPRC
GO:0072503	Cellular divalent inorganic cation homeostasis	0.013		5	CCL1, CCL23, MCOLN1, PLA2G6, PTPRC
GO:0055074	Calcium ion homeostasis	0.011		5	CCL1, CCL23, MCOLN1, PLA2G6, PTPRC
GO:0006874	Cellular calcium ion homeostasis	0.010		5	CCL1, CCL23, MCOLN1, PLA2G6, PTPRC

Table 2. Information of gene ontology (GO) network analysis of genes related to non-synonymous SNPs in Landrace selective sweep regions

SNPs, single nucleotide polymorphisms.

Significant results of GO analysis using genes related to non-synonymous SNPs in the selective sweep regions of the Landrace genome with our criteria in ClueGO packages of Cytoscape (number of genes = 4, sharing group percentage = 40.0). These results are largely divided into eight clusters as follows.

Predicting strong effects of nsSNPs on amino acid substitutions in Landrace selective sweep region

Two in silico SNP prediction algorithms, SIFT [22] and Poly-Phen-2 [23], were applied to estimate the possible effects of the stabilizing residues on protein functions for 345 nsSNPs in Landrace selective sweep regions. The results of SIFT and Polyphen-2 for 345 non-synonymous SNPs are shown in Tables 3, 4.

According to the SIFT analysis, 75 of 345 nsSNPs were classified as being deleterious (for some SNPs, there was low

Table 3. Summary of non-synonymous single amino acid variation in genes of Landrace selective sweep using SIFT and Polyphen-2

		Polyphen-2								
		Benign	Possibly damaging	Probably damaging	Total					
SIFT	Deleterious	29	19	27	75					
	Tolerated	234	21	15	270					
	Total	263	40	42	345					

SIFT, sorting intolerant from tolerant; Polyphen-2, polymorphism phenotyping v2.

Table 4. Fo	orty-six non-synonym	ous SNPs with stron	g effects on	protein functions	based on SIFT and Polyphen-2
			- 4		

SNP	CHR	POS	A1	A2	SIFT prediction	SIFT score	Polyphen-2 prediction	Polyphen-2 score	Gene	Selective sweep
rs328613228	2	16,206,079	Т	G	deleterious	0	probably damaging	0.997	FAM180B	2:16111708:16299440
2:62624837	2	62,624,837	G	А	deleterious	0.017	possibly damaging	0.853	ENSSSCG00000013821	2:62355986:62756249
rs340857214	2	62,625,107	G	А	deleterious	0.021	possibly damaging	0.539		
2:62625190	2	62,625,190	А	Т	deleterious	0.028	possibly damaging	0.934		
rs335820735	2	62,644,986	А	Т	deleterious	0.008	probably damaging	0.999	ENSSSCG00000013822	
rs343007761	2	62,645,014	Т	G	deleterious	0.018	possibly damaging	0.506		
2:62645060	2	62,645,060	Α	G	deleterious	0.012	possibly damaging	0.604		
rs325197977	2	62,645,081	А	G	deleterious	0	possibly damaging	0.934		
2:62669920	2	62,669,920	G	А	deleterious	0.008	possibly damaging	0.934	ENSSSCG00000013819	
2:62669953	2	62,669,953	Т	G	deleterious	0.007	possibly damaging	0.934		
2:62670031	2	62,670,031	G	А	deleterious	0.012	probably damaging	0.999		
rs342394815	2	85,732,226	Т	С	deleterious	0.002	probably damaging	0.999	ENSSSCG00000014078	2:85467258:86506548
rs337260402	2	85,732,237	Т	G	deleterious	0.003	probably damaging	0.97		
rs326720643	2	85,775,718	А	G	deleterious	0.007	probably damaging	0.984	ANKRD31	
rs318473425	2	86,321,677	Т	А	deleterious	0.033	probably damaging	0.995	ANKDD1B	
rs329106718	5	66,654,214	С	Т	deleterious	0	probably damaging	0.993	TAPBPL	5:66396846:66725591
rs326638161	6	43,729,346	Т	С	deleterious	0.007	probably damaging	0.988	DMBT1	6:43719388:43757067
rs322198139	6	43,750,820	G	Т	deleterious	0.017	possibly damaging	0.915		
rs321057648	6	43,750,963	А	G	deleterious	0.009	possibly damaging	0.663		
6:119199835	6	119,199,835	Т	А	deleterious	0.006	probably damaging	0.998	ENSSSCG00000027618	6:119198939:119344591
rs327779736	8	35,181,016	А	Т	deleterious	0	possibly damaging	0.944	ATP8A1	8:34998191:35275833
rs81399633	8	35,181,037	А	G	deleterious	0.023	possibly damaging	0.896		
rs343636299	9	2,311,094	Т	С	deleterious	0.042	probably damaging	1	OVCH2	9:2223331:2577505
rs318298009	9	3,930,944	Т	А	deleterious	0.006	probably damaging	0.996	TRIM3	9:3927497:3978728
9:11129485	9	11,129,485	Т	G	deleterious	0.035	probably damaging	0.995	MOGAT2	9:11120076:11136889
rs340556206	9	11,129,936	Т	С	deleterious	0.013	probably damaging	0.999		
rs81509118	9	11,130,742	А	G	deleterious	0.036	probably damaging	1		
rs342457070	9	11,130,778	С	А	deleterious	0.005	probably damaging	0.991		
rs327337551	9	11,130,783	G	С	deleterious	0.047	possibly damaging	0.697		
rs338381437	9	11,666,878	G	А	deleterious	0.003	probably damaging	0.983	THAP12	9:11449284:11760977
rs81214615	9	41,047,573	Т	А	deleterious	0.024	probably damaging	0.99	KDELC2	9:40793693:41170478
rs339385194	9	41,076,701	G	Т	deleterious	0.04	probably damaging	0.999	EXPH5	
9:56962342	9	56,962,342	А	G	deleterious	0.028	possibly damaging	0.616	ENSSSCG00000026119	9:56869539:57122277
9:56962578	9	56,962,578	А	С	deleterious	0.026	probably damaging	0.994		
rs328160175	9	56,971,732	G	А	deleterious	0.016	probably damaging	0.994	ENSSSCG00000015182	
rs335643554	9	56,980,378	С	Т	deleterious	0.032	possibly damaging	0.539	ENSSSCG00000028463	
rs331490061	9	56,981,034	А	G	deleterious	0.004	possibly damaging	0.927		
rs326014276	10	63,681,709	G	С	deleterious	0.037	possibly damaging	0.944	GJD4	10:63669866:63725092
rs339353031	11	78,365,823	G	А	deleterious	0.008	probably damaging	0.983	CCDC168	11:78318648:78678168
11:78367889	11	78,367,889	G	А	deleterious	0	probably damaging	0.993		
rs342686832	11	78,367,955	А	G	deleterious	0.034	possibly damaging	0.94		
rs325650226	12	15,917,860	Т	С	deleterious	0.002	probably damaging	0.999	MARCH10	12:15890650:15938045
rs336224471	12	15,917,910	А	С	deleterious	0.03	possibly damaging	0.82		
15:55400479	15	55,400,479	А	G	deleterious	0.032	probably damaging	1	WHSC1L1	15:55142754:55608192
rs339461760	16	46,612,542	С	G	deleterious	0.007	probably damaging	0.998	CWC27	16:46472193:46771773
rs324424231	17	46,357,195	А	G	deleterious	0	probably damaging	0.998	ENSSSCG0000007337	17:46275105:46424519

SNPs, single nucleotide polymorphisms; SIFT, sorting intolerant from tolerant; Polyphen-2, polymorphism phenotyping v2. We identified that 46 of 345 non-synonymous SNPs in the selective sweep regions of the Landrace genome had strong effects on protein function as determined with both in silico tools: SIFT and PolyPhen-2.

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confidence in the findings regarding deleteriousness). Poly-Phen-2 calculates the true-positive rate as a fraction of predicted mutations; its results showed that 82 amino acid variants involving nsSNPs in the selective sweep regions of the Landrace genome were likely to exert deleterious functional effects. In addition, 46 of these nsSNPs overlapped with the SIFT results. From the results of the two bioinformatics tools, we reasoned that 46 of the 345 nsSNPs might have strong effects on biological mechanisms during the process of Landrace domestication (Table 4). Forty-six nsSNPs that had strong effects on protein function were distributed among 26 genes and 19 selective sweep regions. In addition, 2:62355986-62756249 among the 55 selective sweep regions containing nsSNPs had the most nsSNPs (37 SNPs), and the results of the two tools for predicting the nsSNP effects showed that 10 of 37 SNPs in 2:62355986-62756249 had strong effects on protein function. This was the largest number of nsSNPs with a strong effect among the total of 55 selective sweep regions of the Landrace genome containing an nsSNP. In addition, three genes belonged to this selective sweep region: ENSSSCG00000013821, ENSSSCG00000013822, and ENSSSCG00000013819. Because the selective region (2: 62355986-62756249) where this gene is located has not been annotated, we estimated the approximate functions of these three genes by analyzing their orthologs. We searched for orthologous genes of these three genes for which the detailed function had been discovered in placental mammals; there were no one-to-one orthologous genes and only many-to-many orthologous genes (Table 5). Because the lists of orthologs of the three genes were the same, we guessed that the functions of the three genes would be very similar. Because the orthologous genes consisted of 18 genes from 8 species from placental mammals and all 18 genes were related to olfactory receptors, we assumed that ENSSSCG00000013821, ENSSSCG 00000013822, and ENSSSCG00000013819 were inferred as olfactory receptors. In a previous study of pig evolution, one of the several significant features of porcine genome expansion involved the olfactory receptor gene family [30]. Martien et al [26] reported that there are 1,301 porcine olfactory receptor genes and 343 partial olfactory receptor genes. This large number of functional olfactory receptor genes most probably reflects the strong reliance of pigs on their sense of smell while scavenging for food. The presence of greater number of nsSNPs in genes related to olfactory receptors suggested important roles of these genes during selection. Additionally, the monoacylglycerol O-acyltransferase 2 (MOGAT2) gene was shown to have the greatest number of nsSNPs with a strong effect among the 90 genes. Five SNPs of the total of 11 nsSNPs in the MOGAT2 gene had strong effects on protein function in this study. Although our GO network analysis did not reveal any particularly important network of MOGAT2, this gene has been reported to be important in porcine backfat adipose tissue, which is related to the concentration of lipid

and lipid synthesis, as revealed by a transcriptome analysis comparing Landrace and other breeds [31]. In addition, 3 of 26 nsSNPs in the *DMBT1* gene were considered to have strong effects on protein function, as revealed by the SIFT and Polyphen-2 results.

DISCUSSION

Given the interest of the meat production industry in improving the meat quality or piglet number, a genetic investigation focusing on the selective sweep regions of the Landrace genome was previously performed [7]. This study provided vital information for domestic pig breeding. In most selective sweep studies using whole-genome sequencing data, all SNPs, including nsSNPs, were used to detect selective sweep regions. As nsSNPs are mutations that alter the amino acid sequences of encoded proteins, their presence results in a phenotypic change in the organism. Such changes are usually subjected to natural selection. In the case of Landrace, the domestication process had a shorter generation interval than natural selection. Therefore, we believe that nsSNPs had a diverse evolutionary history during the domestication and artificial selection processes, and advanced studies are required to achieve an accurate interpretation of the Landrace genome using nsSNP information after exploring Landrace positive selection based on whole-genome sequence data. In this study, we performed several analyses of nsSNPs of the Landrace genome to obtain a better understanding of the whole genome. We assumed that the information on these nsSNPs might be associated with novel important biological mechanisms related to particular traits of the Landrace breed. For the precise analysis of the characteristics of the Landrace breed from a genomic perspective, we investigated the biological meaning of nsSNPs in the selective sweep regions of the Landrace genome used in a previous study [7]. As a result, there was no correlation between the number of nsSNPs and gene length per 90 genes containing an nsSNP within the selective sweep regions of the Landrace genome (Figure 5), which was contrary to our expectations. Considering that 22 of 90 genes overlapped with multiple selective sweep regions while the others belonged to a single selective sweep region, we assumed that genes containing many nsSNPs in the selective sweep regions of the Landrace genome were more meaningful than our expectation. Subsequently, based on GO network analysis using genes containing 345 nsSNPs in the selective sweep regions of the Landrace genome, a large proportion of selective sweep regions of the Landrace genome where strong amino acid sequence changes had occurred, were involved in the superior reproductive capacity or growth and development of the Landrace breed during the perinatal period. Some of the GO network results overlapped with the GO analysis of all the selective sweep regions in a previous study, while others involved novel interpretations
 Table 5. Information on the orthologs of three genes (ENSSSCG00000013821, ENSSSCG00000013822, and ENSSSCG000000138149) in selective sweep 2:62355986-62756249

	Match	Match an and bla		ENSSSCG0000013821			ENSS	SCG00000	013822	ENSSSCG0000013819		
Species	symbol	gene ID	Compare regions	dN/dS	Target %id	Query %id	dN/dS	Target %id	Query %id	dN/dS	Target %id	Query %id
Chimpanzee (Pan troglodytes)	OR7A5	ENSPTRG00000010603	19:15,130,772-15,137,945	0.350	69.0	70.7	0.372	69.6	71.8	0.327	71.2	70.9
Chimpanzee (Pan troglodytes)	OR7A10	ENSPTRG00000010604	19:15,143,753-15,144,682	0.377	70.6	70.1	0.333	71.8	71.8	0.338	71.8	69.4
Gibbon (Nomascus leucogenys)	OR7A17	ENSNLEG00000005159	GL397382.1:231,228-275,098	0.383	71.0	70.7	0.359	70.3	70.6	0.290	73.2	70.9
Gorilla (Gorilla gorilla gorilla)	OR7A10	ENSGGOG0000015049	19:15,120,105-15,121,034	-	70.6	70.1		70.9	70.9		72.2	69.7
Gorilla (Gorilla gorilla gorilla)	OR7A17	ENSGGOG0000034834	19:15,160,189-15,161,115	-	72.5	72.0	-	72.8	72.8	-	73.1	70.6
Human (Homo sapiens)	OR7A10	ENSG00000127515	19:14,840,948-14,841,877	0.418	70.2	69.8	0.377	70.6	70.6	0.361	71.8	69.4
Human (Homo sapiens)	OR7A17	ENSG00000185385	19:14,880,426-14,881,452	0.338	72.2	71.7	0.356	72.5	72.5	0.317	72.5	70.0
Human (Homo sapiens)	OR7A5	ENSG00000188269	19:14,792,490-14,835,376	0.354	69.6	71.4	0.370	70.2	72.5	0.313	71.5	71.3
Mouse (Mus musculus)	Olfr1353	ENSMUSG00000042774	10:78,963,309-78,971,338	-	62.5	62.1	-	61.2	61.2	0.243	65.1	62.8
Mouse (Mus musculus)	Olfr1352	ENSMUSG00000046493	10:78,981,050-78,987,903	0.238	68.6	68.2	0.224	67.3	67.3	-	68.6	66.3
Mouse (Mus musculus)	Olfr19	ENSMUSG00000048101	16:16,672,228-16,676,405	0.245	68.3	67.9	0.267	66.3	66.3	0.253	67.6	65.3
Mouse (Mus musculus)	Olfr57	ENSMUSG0000060205	10:79,028,741-79,036,274	0.308	66.5	68.2	0.289	64.3	66.3	0.349	65.2	65.0
Mouse (Mus musculus)	Olfr1351	ENSMUSG0000063216	10:79,012,472-79,019,645	0.308	64.6	66.2	0.303	62.1	64.1	0.345	64.3	64.1
Mouse (Mus musculus)	Olfr8	ENSMUSG0000094080	10:78,950,636-78,958,378	0.284	63.2	63.0	0.317	58.4	58.6	-	60.7	58.8
Mouse (Mus musculus)	Olfr1354	ENSMUSG0000094673	10:78,913,171-78,920,399	0.264	63.6	63.3		59.0	59.2		62.3	60.3
Orangutan (Pongo abelii)	OR7A5	ENSPPYG00000009655	19:15,004,902-15,005,858	0.373	67.9	69.5	0.395	67.3	69.3	0.351	68.9	68.4
Orangutan (Pongo abelii)	OR7A10	ENSPPYG00000009656	19:15,019,264-15,020,193	0.453	69.6	69.1	0.402	70.9	70.9	0.350	71.2	68.8
Orangutan (Pongo abelii)	OR7A17	ENSPPYG00000009658	19:15,062,903-15,091,843	0.344	70.9	70.4	0.339	71.5	71.5	0.342	70.9	68.4
Rat (Rattus norvegicus)	Olr1073	ENSRNOG0000031688	7:13,378,338-13,379,273	-	62.1	62.1	-	61.7	62.1	0.270	65.3	63.4
Rat (Rattus norvegicus)	Olr1076	ENSRNOG0000039448	7:13,424,355-13,425,311	0.263	66.0	67.5	0.248	63.8	65.7	0.285	64.8	64.4
Rat (Rattus norvegicus)	Olr1075	ENSRNOG0000039449	7:13,403,899-13,404,858	0.290	67.1	68.8	0.272	66.1	68.3	0.291	67.4	67.2
Rat (Rattus norvegicus)	Olr1085	ENSRNOG0000047090	7:13,673,934-13,674,866	-	63.2	63.0	0.343	58.4	58.6	0.327	62.3	60.3
Rat (Rattus norvegicus)	Olr1079	ENSRNOG0000049781	7:13,488,205-13,489,137	0.276	63.6	63.3	0.395	59.4	59.6	0.336	62.6	60.6
Rat (Rattus norvegicus)	Olr1077	ENSRNOG0000054107	7:13,460,476-13,461,405	0.229	69.3	68.8	0.241	67.0	67.0	0.236	68.0	65.6
Rat (Rattus norvegicus)	Olr1082	ENSRNOG0000058943	7:13,553,010-13,553,963	0.279	61.8	63.0	0.348	58.0	59.6	0.342	59.6	59.1
Rat (Rattus norvegicus)	Olr1083	ENSRNOG0000061480	7:13,587,479-13,588,411	0.290	63.2	63.0	0.352	60.3	60.5	0.332	62.6	60.6
Vervet-AGM (Chlorocebus sabaeus)	OR7A10	ENSCSAG00000006193	6:13,469,888-13,471,167	0.347	70.2	69.8	0.330	72.2	72.2	0.348	71.8	69.4

of the Landrace genome [7].

CONCLUSION

Our results strongly suggested that Landrace genetic variants, which could give rise to changes in amino acid sequences, might be important factors for the superior reproductive capacity of this breed. We aimed to perform analyses of the Landrace genome using nsSNPs in selective sweep regions. Our results showed that most of the genes affected by nsSNPs in the selective sweep regions may be closely related to the superior reproductive capacity or growth and development of the Landrace breed during the perinatal period. Furthermore, there were indications that nsSNPs in selection had impacted in Landrace breed establishment. This study will provide insights into the impact of the process of domestication on the Landrace genome.

CONFLICT OF INTEREST

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

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