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Molecular genetic analysis of boar taint

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1 Introduction

1.1 Problem/Knowledge

Intact boars are rarely used for fattening, because consumers would object to the boar taint, which tends to develop with sexual maturity and renders pork inedible. To eliminate this problem, boars are usually castrated at a young age, a practice which is painful and has been criticized repeatedly as not in line with animal welfare. In 2008, representatives of the German pig farming community, the processing industry and the trade drafted a resolution ("Düsseldorfer Erklärung") to stop castration of piglets without anesthezation. European pig farmers and their union (COPA-COGECA) agreed in December 2010 to terminate surgical castration by 2018. This means that castration of piglets with anesthesia will only be accepted as a transitional step until castration will be completely banned in Europe. However, if intact boars are fattened, negative consumer response to boar taint in pork has to be prevented: by testing carcasses routinely with sufficient speed and accuracy and by reducing the incidence of boar taint at slaughter age. This may be approached in different ways: by genetic selection, nutrition and/or management.

Boar taint develops under the influence of genetic and non-genetic factors (Bracher-Jakob, 2000). Several studies have shown that the level of skatole and androstenone, the two main components responsible for boar taint, is moderately to highly heritable; the deposition in fat increases with sexual maturity. Non-genetic contributing factors which have been identified are group *vs.* single pen management and light for *androstenone* level and nutrition, housing system and hygiene for *skatole*.

In order to assess the chances to reduce and eventually eliminate the boar taint by genetic selection, we need to know the relevant population parameters. These estimates should not be taken at face value without taking all essential factors into account: age and live weight at the time of testing, management conditions, laboratory techniques applied, and sample size. As pointed out by Haugen (2010), neither are official reference methods available to determine and compare androstenone and skatole levels, nor are all results being published.

The relevance of laboratory techniques has been demonstrated by Harlizius et al. (2008), whose results from different laboratory methods differed by a factor of 2 to 4 for identical samples of backfat. This should be kept in mind; for genetic evaluation, genotypes must always be compared under the same conditions.

A number of quantitative trait loci (QTL) and genome-wide association analysis have been conducted for androstenone in the purebred and crossbred pig populations (Duijvesteijn et al., 2010; Gregersen et al., 2012; Grindflek et al., 2011; Lee et al., 2004; Quintanilla et al., 2003;

Robic et al., 2011). Gene expression analysis has been used to identify candidate genes related to the trait of interest. Several candidate genes have been proposed for divergent androstenone levels in different pig populations by global transcriptome analysis in boar testis and liver samples (Leung et al., 2010; Moe et al., 2008; Moe et al., 2007). Functional genomics provides an insight into the molecular processes underlying phenotypic differences (Ponsuksili et al., 2011). RNA-Seq is a recently developed next generation sequencing technology for transcriptome profiling that boosts identification of novel and low abundant transcripts (Wang et al., 2009). RNA-Seq also provides evidence for identification of splicing events, polymorphisms, and different family isoforms of transcripts (Marguerat and Bahler, 2010).

1.2 Objectives

The aim of this study was the identification of genes and pathways influencing boar taint and involved in androstenone and skatol metabolism. Therefore polymorphisms in relevant genes were identified and transcriptome analysis using Affymetrix-Chips and RNA-Seq in the two major organs, testis and the liver, involved in androstenone and skatole metabolism was performed.

2 Material und Methods

2.1 Material

2.1.1 Animals and phenotypes

Tissue samples and phenotypes were collected from the Pietrain × F_2 cross and Duroc × F_2 cross animals. F_2 was created by crossing F_1 animals (Leicoma × German Landrace) with Large White pig breed. Fattening performances of each boar was determined on station for 116 days. Animals were slaughtered when on average 90 kg gain was achieved during this test. All the pigs were slaughtered in a commercial abattoir. Carcass and meat quality data were collected according to guidelines of the German performance test (ZDS, 2007). Tissue samples from testis and liver were frozen in liquid nitrogen immediately after slaughter and stored at -80°C until used for RNA extraction. Fat samples were collected from the neck and stored at -20°C until used for androstenone measurements. For the quantification of androstenone an in-house gas-chromatography/mass spectrometry (GC-MS) method was applied as described previously (Fischer et al., 2011). Pigs having a fat androstenone level less than 0.5 µg/g and greater than 1.0 µg/g were defined as low and high androstenone samples, respectively.

2.1.2 DNA and RNA isolation

For the microarray study, 20 animals of 101 crossing boars (Pietran x F2) with high and low androstenone and skatole levels were selected. Average levels of androstenone were at > 470 ng/g fat and of skatole at > 250 ng/g fat.

Based on next generation sequencing techniques ten boars (Duroc x P2) were investigated. These were selected from a pool of 100 pigs and the average androstenone value for these selected animals was $1.36 \pm 0.45 \ \mu g/g$. RNA for RNA-seq was isolated from testis and liver of 5 pigs with extreme high (2.48 \pm 0.56 $\ \mu g/g$) and 5 pigs with extreme low levels of androstenone (0.24 \pm 0.06 $\ \mu g/g$).

In general total RNA was extracted using RNeasy Mini Kit according to manufacturer's recommendations (Qiagen). Total RNA was treated using on-column RNase-Free DNase set (Promega) and quantified using spectrophotometer (NanoDrop, ND8000, Thermo Scientific). RNA quality was assessed using an Agilent 2100 Bioanalyser and RNA Nano 6000 Labchip kit (Agilent Technologies).

For further investigation, selected candidate genes were genotyped in 300 crossing boars (Pietran x F2). Therefore DNA was obtained from muscle tissue using a phenol-chloroform extraction method.

2.2 Methods

2.2.1 Gene expression analysis with Affymetrix chips

Liver gene expressions pattern were produced using 20 GeneChip Porcine Array (Affymetrix). The analysis of microarray raw data was performed with the R software (http://www.r-project.org). For normalization and background correction of the data, the algorithm gcRMA (GeneChip Robust Multichip Average) was used. Carrying out the analysis of expression differences was performed with a linear model for microarray data (limma) (Smyth, 2004). Three comparisons were taken into account by means of linear contrasts: (1) the comparison of high vs. low skatole, (2) high vs. low androstenone and (3) the interaction between skatole and androstenone. Differentially regulated genes were identified on the basis of a p \leq 0.05, one fold changes \geq 1 and a false discovery rate (FDR) \leq 0.3. The functional annotation of differentially expressed genes was performed by the DAVID (The Database for Integrated Annotation, Visualization and Discovery) gene annotation tool (http://david.abcc.ncifcrf.gov/).

2.2.2 Gene expression analysis with RNA-Seq

Library construction and sequencing

Full-length cDNA was obtained from 1 µg of RNA, with the SMART cDNA Library Construction Kit (Clontech, USA), according to the manufacturer's instructions. Libraries of amplified RNA for each sample were prepared following the Illumina mRNA-Seq protocol. The library preparations were sequenced on an Illumina HiSeq 2000 as single-reads to 100 bp using 1 lane per sample on the same flow-cell (first sequencing run) at GATC Biotech AG (Konstanz, Germany). All sequences were analysed using the CASAVA v1.7 (Illumina, USA).

Reference sequences and alignment

Two different reference sequence sets were generated from NCBI Sscrofa 9.2 assembly. (1) The reference sequence set generated for differential expression analysis comprised of RefSeq mRNA sequences (cDNA sequences) and candidate transcripts from NCBI UniGene database (Sscrofa). (2) For gene variation analysis a different reference sequence set, generated from whole genome sequence (chromosome assembly) was used. During sequencing experiment Sscrofa NCBI 10.2 assembly was not released and Sscrofa 9.2 covered ~8.5 K unannotated SNPs (dbSNP database). The released Sscrofa 10.2 assembly consists of 566 K SNP (http://www.ncbi.nlm.nih.gov/Taxonomy/Browser/wwwtax.cgi?mode=Info&id=9823) with annotation information for 460 K SNP (dbSNP database). In order to make use of this **SNP** NCBI enriched information. we used Remap tool (http://www.ncbi.nlm.nih.gov/genome/tools/remap) to convert Sscrofa 10.2 SNP genomic positions to Sscrofa9.2 positions. Raw reads were mapped to reference sets using BWA algorithm (http://bio-bwa.sourceforge.net/) with the default parameters (Li and Durbin, 2009).

Differential gene expression analysis

For differential gene expression analysis with raw count data a R package DESeq was used (Anders and Huber, 2010). To model the null distribution of the count data, DEseq follows an error model that uses the negative binomial distribution, with variance and mean linked by local regression. The method controls type-I error and provides good detection power (Anders and Huber, 2010). After analysis using DESeq, DEGs were filtered based on *p*-adjusted value (Benjamini and Hochberg, 1995) 0.05 and fold change > 1.5.

Gene variation analysis

For gene variation analysis the mapping files generated by aligning the raw reads to reference sequence set (2) were used. All the downstream analysis was performed using Genome 2010) Analysis Toolkit (GATK) (McKenna et al., and Picard Tools (http://picard.sourceforge.net/). The Genome Analysis Toolkit (GATK) was used for local realignment incorporating Sscrofa 9.2 converted SNPs which was described in the previous section. Covariate counting and base quality score recalibration were done using the default parameters suggested by GATK toolkit. The re-aligned and recalibrated mapping files were grouped according to tissue and phenotype categories. Variant calling was performed for each group using GATK UnifiedGenotyper (McKenna et al., 2010). All the variant calls with a read coverage depth < 75 and base quality < 20 were discarded from further analysis. Polymorphisms identified in DEGs are given in the results section.

Pathways and networks analysis

A list of the DEGs was uploaded into the Ingenuity Pathway Analysis (IPA) software (Ingenuity Systems, www.ingenuity.com) to identify relationships between the genes of interest and to uncover common processes and pathways. Networks of the genes were then algorithmically generated based on their connectivity. The 'Functional Analysis' tool of the IPA software was used to identify biological functions which were most significant to the data set. Canonical pathway analysis was also utilized to identify the pathways from the IPA library of canonical pathways that were most significant to the data set. Fisher's exact test was used to calculate a *p*-value determining the probability that each biological function or canonical pathway assigned to the data set. In addition, the significance of the association between the data set and the canonical pathway was calculated as the ratio of the number of genes from the data set that mapped to the pathway.

Quantitative real-time PCR (qRT-PCR) analysis

Total RNA from testis and liver was isolated from 10 boars for qRT-PCR experiment. cDNA were synthesised by reverse transcription PCR using 2 µg of total RNA, SuperScript II reverse transcriptase (Invitrogen) and oligo(dT)12 primer (Invitrogen). Gene specific primers for the qRT-PCR were designed by using the Primer3 software (Rozen and Skaletsky, 2000). Detailed information for primers used in this study was given in Table 1. Nine-fold serial dilution of plasmids DNA was prepared and used as a template for the generation of the standard curve. In each run, the 96-well microtiter plate contained each cDNA sample, plasmid standards for the standard curves and no-template control. For each PCR reaction 10 µl iTaqTM SYBR® Green Supermix with Rox PCR core reagents (Bio-Rad), 2 µl of cDNA $(50 \text{ ng/}\mu\text{l})$ and an optimized amount of primers were mixed with ddH₂O to a final reaction volume of 20 µl per well. The qRT-PCR was conducted with the following program: 95 °C for 3 min and 40 cycles 95 °C for 15 s/60 °C for 45 s on the StepOne Plus qPCR system (Applied Biosystem). As a technical replication, all samples were repeated and the mean of the two replications was finally used. Final results were reported as the relative expression level compared after normalization of the transcript level using two housekeeping genes PPIA and GAPDH.

Gene	Reference ID	Primer sequences $(5' \rightarrow 3')$	Annealing temperature (°C)	Product size (bp)
UCHL1	NM_213763	F: AGCTGTCGATGGAGCAAGTT	55	136
	NR 001027150	R: CCACATCCAAAGGCCTTAAA F: GTTTGCATCTTGGGGACACT	~~	144
SULIZAI	NM_001037130	R:ATGGGAACAGCTCTTGAGGA	55	144
ESR1	EF195769	F: AGCACCCTGAAGTCTCTGGA	60	146
20111	21 1/07 0/	R:GACAGGATGAGGAGGAGCTG		110
GSTA2	NM_213850	F: TGTTGAAGAGCCATGGACAA	55	131
		R: CTTCAGCAGAGGGAAGTTGG		
DHRS4	NM_214019	F:TCCTGATGACAAAGGCAGTG R:TGCCTTATCCATCCACAACA	60	148
CVD2C22	NM_214414	F: AGCTGTGCCTCATCCCTAGA	56	133
CYP2C33		R: GTGTTTCTGTCCCAGGCAAT	50	
TNC	NM_214230	F:GTGACGGAAGAAACCGTAA	59	119
		R: CTCCAGGGACTCTGAACTGC	57	-
HSD3B1 NN	NM 001004049	F:TCCCCAGTGTTTTCTGGTTC	55	135
		R:CCTTCTCCTCCAGCAACAAG		
HSD17B2	NM_001167649	F: TGCAGAACAGAGGACTGTGG	54	103
		R: GCCATGCATCGTTTGTATTG		
IGFBP1	NM_001195105	F: CCTGCCAGCGAGAACTCTAC	58	131
		R: CTCGCACTGTTTGCTGTGAT		
CYP7A1	NM_001005352		54	104
PLIN2	NM_21400	R-TCTCAGGTTGCTGGGTCTCT	60	102
		F·GGCCTGAAGCCTAAACACAG		
FMO5	XM_001928594	R:CCTGGAGCCATCCTCAAATA	55	147
		F: CACAAACGGTTCCCAGTTT		
PPIA	NM_214353	R: TGTCCACAGTCAGCAATGGT	58	171
GAPDH	AF017079	F:ACCCAGAAGACTGTGGATGG R:ACGCCTGCTTCACCACCTTC	60	247

Table 1Details of primers used for qRT-PCR analysis

2.2.3 SNP-genotyping for the association study and statistical analysis

To identify polymorphisms within candidate genes, specific primers were designed based on published sequences by using Primer3 software (Rozen and Skaletsky, 2000). A list of primers used in this study is given in Table 2.

Gene	Primer pairs	Frag- ment (bp.)	Tm (°C)	Enzyme	SNP position
FMO1	Fw:5′TTCGGGTCCTGAAAGGTAAA-3′ Rv: 5′-TGTGCTGGTAATGGCACAAA-3′	240	55	BtsCI	Intron 7 g.256, A>C
FMO5	Fw: 5'-AATTCTGCACATTCCCCTGA-3' Rv: 5'-CCTGTTTGTTTCCTTGATTGC-3'	223	55	BtsCI	Intron4 g.494, A>G
ESR1	Fw: 5′- GTTCAAATCCCTGGTTGCAT-3′ Rv: 5′-CTAGGCGTCTCCCCAGATTAG-3′	305	60	BstNI	Exon 1 g.672, C>T
CYP21	Fw: 5′-GGTAACCTGTCCCCTCCTG-3′ Rv: 5′-GGTAAGAGACGGCACAGGAG-3′	247	59	Dra III	3'UTR g.3911,T> C
PLIN2	Fw: 5'-TCAAGGCACTCAGGATAAGC-3' Rv: 5'-GAACACTGAGGAGCCTGGTA-3'	196	55	BsmI	Exon 7 g.183/184, G>A
PLIN22	Fw: 5′- TCAAGGCACTCAGGATAAGC-3′ Rv: 5′- GAACACTGAGGAGCCTGGTA-3′	176	55	BsTUI	Exon 8 g.198, G>A

 Table 2
 Polymerase chain reaction primers used for SNPs screening

Polymerase Chain Reactions (PCR)

Polymerase Chain Reactions were performed in a 20 μ l volume containing 2 μ l of genomic DNA, 10×PCR buffer (with 2.0 μ l MgCl2), 1.0 μ l of dNTP, 0.5 μ l of each primer and 0.2 μ l of Taq DNA polymerase (GeneCraft). The PCR were performed under the following

condition: initial denaturing at 95 °C for 5 min followed by 35 cycles of 30 sec at 95 °C, 30 sec at respective annealing temperatures (as given in Table 4) and 10 sec at 72 °C and a final elongation of 10 min at 72 °C.

Genotyping

The PCR-RFLP method was used for genotyping the boars. The restriction enzymes were selected according to the recognition (http://tools.neb.com/NEBcutter2/index.php) of the polymorphic sites. The fragments with the detected mutation were amplified using different annealing temperatures to get the the PCR products (Table 2).An aliquot of the PCR product of each reaction was checked on 1.5% agarose gel (Fisher Scientific Ltd.) before digestion using different endonucleases. The digested products were separated using 2.0% agarose gel. The fragments were visualised under ultraviolet light, and the sizes and the number of fragments analysed using the molecular analyst software (Bio-Rad Laboratories, Molecular Bioscience Group).

Statistical Analysis of the association study

Allele and genotype frequencies of each population were determined to detect SNP in the six candidate genes. The association of the genotypes from six candidate genes with boar taint compounds were calculated by analyzing variance of quantitative traits. For these analyses a generalized linear model of SAS (SAS Inst. Inc., Cary, NC) was used. The model was as follows:

 $Y_{ijklm} = \mu + season_i + genotype_j + station_k + pen_l + e_{ijkl}$

Where Y is the boar taint compounds (Skatole, Androstenone and Indole), μ is overall mean, season is the fixed effect of i-th season (i= winter/summer), genotype is the fixed effect of j-th genotype (j=1,2, and 3), station is the fixed effect of k-th station (Grub, Schwarzenau, Frankenforst, Haus Düsse and Boxberg), pen is the fixed effect of l-th pen (group, individual), and e_{ijkl} is the residual error.

The distribution of the genotypes and accuracy of genotype scoring was tested for Hardy–Weinberg equilibrium by chi-square (X^2) test before using both polymorphisms for the association analysis.

3 Results

3.1 Gene expression analysis with Affymetrix chips

Differentially regulated genes based on the comparison of high vs. low skatole and high vs. low androstenone are described in Table 3 and Table 4. Generally 107 genes were differentially expressed comparing high and low skatole. 49 were up regulated and 58 were down regulated. The investigation of differentially expressed genes related to a divergent andostrenone level revealed only two genes (Figure 1).



Figure 1 Venn diagram of differentially expressed genes in the different analysed groups

A gene ontology classification was performed using the online tool DAVID in order to assign differentially expressed genes to categories biological functions and pathways. Differentially expressed genes between the respective groups showed significant features in catalytic activities, metabolic processes, fatty acid metabolism and lipid metabolic processes.

Investigating the data using an interaction term between skatole and andostenone revealed a different set of differentially expressed genes. The gene FMO1 (*Flavin containing monooxygenase 1*) was identified within this step, and seems to be promising, because it is involved in the phase I metabolism of skatole and andostenone.

Gene symbol	Gene name	logFC	p-value	FDR	
O5TCH4	cytochrome P450, family 4, subfamily A,	-3 92794	0.000214	0 069048	
Q310114	polypeptide 22	5.72774	0.000214	0.007070	
SCD	Acyl-CoA desaturase	-3.54469	0.000695	0.121177	
FASN	Fatty acid synthase	-3.24506	2.78E-05	0.027319	
CYP4A11	Cytochrome P450 4A11	-3.11157	0.000108	0.050052	
PSPHL	L-3-phosphoserine phosphatase	-2.59417	3.89E-05	0.028635	
PPAP2A	Lipid phosphate phosphohydrolase 1	-2.55253	0.002424	0.20664	
ACAS2	Acetyl-coenzyme A synthetase, cytoplasmic	-2.39256	0.000789	0.127635	
DOCK1	Dedicator of cytokinesis protein 1	-2.1362	2.02E-05	0.027021	
PTPRD	Protein-tyrosine phosphatase delta precursor	-2.05933	0.000369	0.089305	
FDPS	Farnesyl pyrophosphate synthetase	-2.04901	0.000132	0.054027	
Q7Z4L5	tetratricopeptide repeat domain 21B	-1.87824	0.001175	0.148852	
ALDH1A1	Aldehyde dehydrogenase 1A1	-1.85114	0.002246	0.202877	
NEIL1	nei endonuclease VIII-like 1; endonuclease VIII	-1.8084	0.000862	0.130409	
NP_056193	vacuolar protein sorting 13D	-1.71193	0.00348	0.23542	
TM7SF2	Delta(14)-sterol reductase	-1.68122	2.91E-05	0.027319	
ACAS2	Acetyl-coenzyme A synthetase, cytoplasmic	-1.67735	0.001359	0.152922	
Q9BRJ6		-1.64633	0.005114	0.271692	
CA7	Carbonic anhydrase VII	-1.58986	0.00145	0.15645	
PPAP2A	Lipid phosphate phosphohydrolase 1	-1.58363	0.000206	0.069048	
Q7Z3I3		-1.58343	0.000385	0.089373	
NP_835229	similar to delta 5 fatty acid desaturase	-1.55658	0.002351	0.204862	
DHCR7	7-dehydrocholesterol reductase	-1.54429	0.00029	0.080099	
ACACA	Acetyl-CoA carboxylase 1	-1.50628	0.005438	0.280892	
CYP2D6	Cytochrome P450 2D6	-1.43153	0.001724	0.171839	
NP_998771	similar monocarboxylate transporter	-1.43106	0.00085	0.129976	
ANK3	Ankyrin 3 (ANK-3)	-1.37404	0.001994	0.187931	
PTPRD	Protein-tyrosine phosphatase delta precursor	-1.32185	0.000228	0.069048	
CCAT	2-amino-3-ketobutyrate coenzyme A ligase,	1 21(02	2.755.05	0.027210	
GCAI	mitochondrial precursor	-1.31093	2.75E-05	0.027319	
NP_054745	heat shock-like protein 1	-1.31089	0.004287	0.254694	
ACY1L2	ACY1L2 protein	-1.29007	0.002486	0.20873	
AGT	Angiotensinogen precursor	-1.27926	0.003591	0.237744	
NEIL1	nei endonuclease VIII-like 1	-1.26648	0.0006	0.110157	
C14orf1	UPF0143 protein C14orf1	-1.26279	0.000211	0.069048	
UGT2B17	UDP-glucuronosyltransferase 2B17 precursor, microsomal	-1.23704	0.002944	0.223935	

 Table 3
 Differentially expressed genes based on microarrays – high skatol group versus low skatol group

Gene symbol	Gene name	logFC	p-value	FDR	
NSDHL	NAD(P)-dependent steroid dehydrogenase	-1.22758	0.001309	0.151111	
RORA	Nuclear receptor ROR-alpha	-1.21621	0.001338	0.152922	
BBX	HMG-BOX transcription factor BBX	-1.20562	0.000212	0.069048	
AFM	Afamin precursor (Alpha-albumin)	-1.16492	0.002925	0.223535	
GSTT1	Glutathione S-transferase theta 1	-1.14228	0.002952	0.223935	
	Complement-activating component of Ra-	1 10161	0.004702	0 265575	
MASPI	reactive factor precursor	-1.10161	0.004703	0.265575	
ALDH1A1	Aldehyde dehydrogenase 1A1	-1.0969	0.00083	0.129457	
NP_835229	similar to delta 5 fatty acid desaturase	-1.09457	0.001922	0.18326	
Q96CU9		-1.06556	0.000482	0.09614	
SLC23A1	Solute carrier family 23, member 1	-1.0447	0.003333	0.231735	
BPHL	Valacyclovir hydrolase precursor	-1.03997	0.003911	0.245289	
HABP2	hyaluronan binding protein 2	-1.02274	0.000713	0.121177	
AGMAT	Agmatinase, mitochondrial precursor	-1.01449	0.000227	0.069048	
DHRS3	Short-chain dehydrogenase/reductase 3	-1.00901	0.001364	0.152922	
CYP39A1	Cytochrome P450 39A1	-1.00717	0.003776	0.241814	
TAX1BP1	Tax1 binding protein	1.01058	3.50E-05	0.027523	
	Group XIIA secretory phospholipase A2	1.010100		0.100155	
PLA2G12A	precursor	1.013123	0.000833	0.129457	
	Trifunctional enzyme alpha subunit,	1.0040(7	0.000000	0.001705	
HADHA	mitochondrial precursor	1.024067	0.003332	0.231735	
C5orf3		1.065691	0.003584	0.237744	
SEC23A	Protein transport protein Sec23A	1.085506	0.001295	0.151008	
NP_060189	SNF-1 related kinase	1.088831	8.63E-05	0.043346	
KIAA1423	PREDICTED: KIAA1423	1.097086	0.001191	0.148856	
TCOL N2	Trans-Golgi network integral membrane protein	1 102042	0.000200	0.00000	
I GOLINZ	2 precursor	1.105042	0.000298	0.080099	
ITGAV	Integrin alpha-V precursor	1.104383	0.004615	0.263806	
DOLPP1	Dolichyldiphosphatase 1	1.120522	0.002331	0.204451	
Q96N33		1.136118	0.001449	0.15645	
C9orf67		1.142002	0.000447	0.091742	
CA00	Protein CGI-100 precursor	1.191502	4.41E-06	0.01368	
GALT	Galactose-1-phosphate uridylyltransferase	1.207998	0.004722	0.265575	
Q8NEX2		1.215337	0.001223	0.149388	
Q96N33		1.217515	0.00074	0.123181	
SEC23A	Protein transport protein Sec23A	1.228197	0.001803	0.174722	
IDH1	Isocitrate dehydrogenase [NADP] cytoplasmic	1.250375	5.13E-05	0.032647	
M6PRBP1	Mannose-6-phosphate receptor binding protein 1	1.273556	0.001777	0.173516	
BR44	Brain protein 44.	1.274216	1.27E-05	0.021906	
ADAMTS19	ADAMTS-19 precursor	1.274353	0.002257	0.203166	

Gene symbol	Gene name	logFC	p-value	FDR
PGM3	Phosphoacetylglucosamine mutase	1.316116	0.001756	0.173516
КСҮ	UMP-CMP kinase	1.332816	0.000302	0.080099
ADR2	Adiponectin receptor protein 2	1.364597	0.00485	0.26765
MARVELD3	MARVEL domain containing 3;	1.389535	0.000506	0.097365
PPARA	Peroxisome proliferator activated receptor alpha (PPAR-alpha).	1.405004	0.004912	0.268148
SLC25A20	Mitochondrial carnitine/acylcarnitine carrier protein	1.419251	0.001241	0.149673
RDH11	Retinol dehydrogenase 11	1.419451	0.002785	0.216843
LCF1	Long-chain-fatty-acidCoA ligase 1	1.419931	2.44E-05	0.027319
F34A	Protein FAM34A. 4]	1.426573	0.002513	0.20873
PRDM6	PR-domain zinc finger protein 6	1.434763	0.005952	0.297893
F34A	Protein FAM34A.	1.441989	0.003532	0.236648
ACADVL	Acyl-CoA dehydrogenase, very-long-chain specific, mitochondrial precursor	1.453334	0.003998	0.248398
F34A	Protein FAM34A.	1.463965	0.005274	0.276589
CIDEB	Cell death activator CIDE-B (Cell death- inducing DFFA-like effector B).	1.473663	0.001585	0.163371
GPD1	Glycerol-3-phosphate dehydrogenase [NAD+], cytoplasmic	1.476243	3.06E-05	0.027319
FBP1	Fructose-1,6-bisphosphatase	1.477433	0.001179	0.148852
BFAR	apoptosis regulator	1.482646	0.003028	0.225327
TM4SF13	Transmembrane 4 superfamily member 13	1.501593	0.0002	0.069048
COQ6	Ubiquinone biosynthesis monooxgenase COQ6	1.5069	0.001301	0.151008
GPD1L	glycerol-3-phosphate dehydrogenase 1-like	1.534862	0.003873	0.244565
NP_951038	I-mfa domain-containing protein isoform p40	1.545929	4.88E-05	0.032647
NP_951038	I-mfa domain-containing protein isoform p40	1.567372	2.55E-05	0.027319
ADR2	Adiponectin receptor protein 2	1.608796	0.003867	0.244565
KCY	UMP-CMP kinase	1.626891	0.000256	0.074283
ARF4	ADP-ribosylation factor 4.	1.640433	0.001482	0.15682
C10orf45		1.646974	0.003455	0.23542
Q68D10		1.809643	0.003599	0.237744
PLA2G7	Platelet-activating factor acetylhydrolase precursor	1.859754	0.000112	0.050092
ABCD3	ATP-binding cassette, sub-family D, member 3	2.004399	1.34E-06	0.01368
DHRS4	Dehydrogenase/reductase SDR family member 4	2.077268	0.003558	0.237108
ABCD3	ATP-binding cassette, sub-family D, member 3	2.178742	3.82E-06	0.01368
LDHB	L-lactate dehydrogenase B chain	2.53541	4.33E-06	0.01368
ABCD3	ATP-binding cassette, sub-family D, member 3	2.843671	1.25E-05	0.021906
G0S2	Putative lymphocyte G0/G1 switch protein 2.	2.976977	0.001957	0.185606

Gene symbol	Gene name	logFC	p-value	FDR	
G0S2	Putative lymphocyte G0/G1 switch protein 2.	3.012107	0.001479	0.15682	
PMM1	Phosphomannomutase 1	3.325454	2.78E-05	0.027319	
ADFP	Adipophilin	3.673963	0.001301	0.151008	
HMGCS2	Hydroxymethylglutaryl-CoA synthase,	4 529646	0.000114	0.050092	
IIWOC52	mitochondrial precursor	4.527040	0.000114		

Table 4	Differentially expressed	genes – hig	gh andostenone	group versu	is low
	androstenone group				
Gene symbol	Gene name		log F	C p-value	FDR
CYP3A7	Cytochrome P450 3A7		1.337	88 0.001722	0.844666
INHBA	Inhibin beta A chain precursor		-1.066	619 0.001891	0.844666

3.2 Analysis of RNA-Seq data

We sequenced cDNA libraries from 10 samples per tissue using Illumina HiSeq 2000. The sequencing produced clusters of sequence reads with maximum 100 base-pair (bp) length. After quality filtering the total number of reads for testis and liver samples ranged from 13.2 million (M) to 33.2 M and 12.1 M to 46.0 M, respectively. There was no significant difference in the number of reads from low and high androstenone samples (p = 0.68). Total number of reads for each tissue group and the number of reads mapped to reference sequences are shown in Table 5 and Table 6. In case of testis 42.20% to 50.34% of total reads were aligned to reference sequence whereas, in case of liver 40.8% to 56.63% were aligned.

Group	Sample	Total number of reads	Un-mapped reads	Mapped reads	Percentage of unmapped reads	Percentage of mapped reads
	A1	15,142,756	7,811,096	7,331,660	51.50	48.50
	A2	13,221,550	6,564,679	6,656,871	49.66	50.34
Low androstenone	A3	32,389,084	16,697,785	15,691,299	51.50	48.50
	A4	27,068,779	14,123,318	12,945,461	52.10	47.90
	A5	27,015,712	14,465,669	12,550,043	53.54	46.46
High androstenone	A6 A7 A8 A9 A10	32,691,057 33,206,723 15,111,453 14,330,069 15,605,400	18,919,738 17,271,473 7,764,418 8,070,092 8,276,052	13,771,319 15,935,250 7,347,035 6,259,977 7,329,348	57.80 51.20 51.38 56.31 53.30	42.20 48.80 48.62 43.69 46.70

Table 5 Summary of sequence read alignments to reference genome in testis samples

Group	Sample	Total number of reads	Un- mapped reads	Mapped reads	Percentage of unmapped reads	Percentage of mapped reads
	B1	29,549,267	15,632,809	13,916,458	53.50	46.50
Law	B2	46,050,468	25,270,695	20,779,773	54.87	45.13
Low	B3	16,420,055	7,659,515	8,760,540	46.64	53.36
androstenone	B4	13,323,763	6,989,584	6,334,179	52.46	47.54
	B5	27,085,837	11,747,225	15,338,612	43.37	56.63
	B6	28,976,693	16,123,777	12,852,916	55.64	44.36
High	B7	12,755,487	5,879,896	6,875,591	46.10	53.90
nigii	B8	45,203,089	18,443,608	26,759,481	59.20	40.8
androstenone	B9	14,559,329	8,540,379	6,018,950	58.66	41.34
	B10	14,527,329	8,062,992	6,464,337	55.51	44.49

Table 6Summary of sequence read alignments to reference genome in liver samples

3.3 Differential gene expression analysis based on RNA-Seq

Differential gene expression for testis and liver with divergent androstenone levels were calculated from the raw reads using the R package DESeq (Anders and Huber, 2010). The significant scores were corrected for multiple testing using Benjamini-Hochberg correction. We used a negative binomial distribution based method implemented in DESeq to identify differentially expressed genes (DEGs) in testis and liver with divergent androstenone levels. A total of 46 and 25 DEGs were selected from the differential expression analysis using the criteria $p_{\text{adjusted}} < 0.05$ and fold change ≥ 1.5 for testis and liver tissues respectively (Table 7 and Table 8). In testis tissues, 14 genes were found to be highly expressed in high androstenone group whereas, 32 genes were found to be highly expressed in low androstenone group. In the liver tissue, 9 genes were found to be highly expressed in high androstenone group whereas, 16 genes were found to be highly expressed in low androstenone group (Table 7 and Table 8). The range of log fold change values for DEGs was from -4.68 to 2.90 for testis and from -2.86 to 3.89 for liver. Heatmaps (Figure 1, A and B) illustrate the DEGs identified in high and low androstenone testis and liver tissues. The differential expression analysis of our data revealed both novel transcripts and common genes which were previously identified in various gene expression studies. Novel transcripts from our analysis and commonly found genes are mentioned in detail in the discussion section.



Figure 2 Heatmap showing differentially expressed genes in (A) testis and (B) liver samples.

The red blocks represent over expressed genes, and the green blocks represent under expressed genes. Legend: A1-A5 testis with low androstenone and A6-A10 testis with high androstenone, B1-B5 liver with low androstenone and B6-B10 liver with high androstenone.

Cana	Deference ID	Log fold	m adjusted
Gene	Reference ID	change	<i>p</i> -adjusted
DKK2	XM_003129269.1	2.89	4.46e-06
AMN	XM_001925648.2	2.28	0.025
LOC100519550	XM_003127761.1	2.22	9.67e-12
CYP4B1	XM_003128017.1	2.20	8.55e-10
CD244	XM_001928325.2	2.15	5.35e-08
ADAMTS4	XM_001927507.2	2.10	0.04
CYP4A11	XM_003128032.1	2.03	3.38e-06
HAL	XM_001925061.1	2.03	0.014
CYP2C33	NM_214414.1	1.91	0.04
AMY2B	XM_003125887.1	1.85	0.002
ARG2	XM_001928679.2	1.84	6.96e-06
LOC100516362	XM_003124870.1	1.74	0.022
LOC100521272	XM_003126855.1	1.71	4.46e-05
MSMO1	NM_213752.1	1.66	5.55e-08
KRT4	XM_001927218.2	-1.52	0.02
MPP7	XM_003130762.1	-1.54	0.0004
DSP	XM_003128168.1	-1.55	6.03e-05
AMHR2	XM_003126187.1	-1.58	0.025
SLA-3	AB105388.1	-1.60	5.15e-07

Table 7 Differentially expressed genes in testis androstenone samples

Cana	Deference ID	Log fold	n-adjusted	
Gelle	Kelelelice ID	change	<i>p</i> -aujusteu	
HAAO	XM_003125193.1	-1.61	0.004	
MX1	NM_214061.1	-1.62	1.63e-09	
MX2	NM_001097416.1	-1.63	7.47e-06	
IFIT2	XM_001928671.2	-1.64	0.0094	
HBB	NM_001144841.1	-1.69	2.12e-08	
ARL4C	XM_003133753.1	-1.72	0.04	
EDN1	NM_213882.1	-1.73	0.004	
HBM	XM_003124683.1	-1.74	0.04	
HBD	XM_003129515.1	-1.83	1.93e-07	
HBA2	XM_003124688.1	-1.90	7.42e-11	
HBA2	XM_003124690.1	-1.90	2.17e-10	
HBA2	XM_003124687.1	-1.93	1.87e-11	
HBA2	XM_003124689.1	-1.95	2.62e-11	
HBA2	XM_003124685.1	-1.97	1.87e-11	
HBA2	XM_003124684.1	-1.97	1.87e-11	
HBA2	XM_003124686.1	-1.99	1.38e-11	
FRK	XM_001925792.2	-2.12	0.002	
IRG6	NM_213817.1	-2.17	7.24e-07	
SYT10	XM_001927016.2	-2.23	9.03e-05	
S100A2	XM_001929559.1	-2.35	0.0008	
CD5	XM_003122679.1	-2.42	0.02	
CYP2B22	NM_214413.1	-2.48	0.02	
CYTL1	XM_003128849.1	-2.82	0.002	
S100A2	XM_001929556.1	-2.83	1.42e-07	
CHRNA3	XM_001925760.2	-3.45	5.07e-08	
OLFRA03	XM_001926523.1	-4.12	0.01	
KRT82	XM_003126157.1	-4.68	1.49e-09	

Cana	Deferrer og ID	Log fold	n adjusted	
Gene	Reference ID	change	<i>p</i> -adjusted	
LOC100512122	XM_003130359.1	3.89	1.10e-14	
LOC100511195	XR_115925.1	3.57	9.26e-15	
IP6K1	XM_001925759.2	3.04	0.002	
AMPD3	XM_003135226.1	2.99	0.0004	
LOC100521668	XR_116002.1	2.52	7.77e-08	
SDS	XM_001928302.2	2.12	8.15e-05	
BTG3	XM_003132741.1	2.12	1.51e-06	
KRT78	XM_001927194.2	2.09	7.80e-05	
SMPDL3A	XM_003121227.1	1.99	9.38e-05	
KRT8	NM_001159615.1	1.96	6.50e-05	
LEAP2	NM_213788.1	1.94	3.84e-06	
HAL	XM_001925061.1	1.91	3.36e-06	
NNMT	NM_001123146.1	1.86	3.83e-05	
BTG3	NM_001097517.1	1.70	0.0007	
KRT18	XM_003126180.1	1.69	0.010	
CDKN1A	XM_001929558.1	1.67	6.50e-05	
TSKU	XM_003129674.1	-1.72	0.0017	
FMO5	XM_001928594.1	-1.75	0.0043	
TSKU	XM_003129672.1	-1.81	0.002	
TSKU	XM_003129673.1	-1.81	0.002	
CYP7A1	NM_001005352.2	-1.87	3.32e-07	
HIST1H4K	XM_001928022.2	-2.60	0.001	
MBL2	NM_214125.1	-2.79	0.0001	
BCAM	XM_003127227.1	-2.83	0.017	
HSD17B2	NM_001167649.1	-2.86	3.92e-09	

Table 8Differentially expressed genes in liver androstenone samples

To investigate gene functions and to uncover the common processes and pathways among the selected DEGs, Ingenuity Pathway Analysis (IPA) software (Ingenuity Systems, www.ingenuity.com) was used. In testis samples, out of 46 DEGs 39 were assigned to a specific functional group based on the information from IPA (Figure 3). A large proportion (84.7%) of the DEGs from testis high androstenone group fell into Gene Ontology (GO) categories such as molecular transport, small molecule biochemistry, amino acid metabolism, embryonic development, carbohydrate metabolism, lipid metabolism and reproductive system development and function (Figure 3).



Figure 3Functional grouping of DEGs in testis with high and low androstenone using
Ingenuity Pathways Analysis (IPA) software.

The most significant functional groups (p < 0.05) are presented graphically. The bars represent the *p*-value on a logarithmic scale for each functional group.

The genes classified into each functional group are listed in the Table 9. The differentially expressed genes MSMO1 and ARG2 are involved in arginine degradation metabolic pathway and additionally, ARG2 is found to be involved in citruline biosynthesis and urea cycle pathways. The gene MSMO1 is also involved in cholesterol biosynthesis and zymosterol synthesis. The differentially expressed cytochrome family gene CYP4A11 is involved in alpha-tocopherol degradation.

IPA assigned 104 DEGs between high and low androstenone testis samples to eleven different canonical pathways. These enriched pathways were metabolic pathways including retinol, trypthopan, arginine and proline, fatty acid and sulphur metabolism (Figure 4). Other pathway categories, including LXR/RXR activation, valine, leucine & isolenone degradation,

biosynthesis of steroid, butanoate, LPS/ILI mediated and IL-10 signaling were also enriched (Figure 4).



Figure 4 The most prominent canonical pathways related to the DEGs data (p < 0.05) for testis with high and low androstenone.

The bars represent the *p*-value for each pathway. The orange irregular line represents the ratio (genes from the data set/total number of genes involved in the pathway) for the different pathways.

For the liver androstenone samples, out of 25 DEGs, 22 could be assigned to a specific functional group based on the information from IPA (Figure 5). A large proportion (88.0%) of the DEGs from liver high androstenone group was enriched with GO functional categories such as amino acid metabolism, small molecule biochemistry, cellular development, lipid metabolism, molecular transport, cellular function and maintenance and cellular growth and proliferation. The genes classified into each group are listed in the Table 10. Among the differentially expressed genes in liver samples, CDKN1A and HSD17B2 are involved in VDR/RXR activation metabolic pathway and CYP7A1 and FMO5 genes are involved in LPS/IL-1 mediated inhibition of RXR function pathway.

Function	Number of	r voluo*	Conos		
Function	genes	<i>p</i> -value*	Genes		
			HBB, HBD, HBA1/HBA2,		
Molecular transport	9	1.00E-05 to 4.96E-02	CYP4A11,EDN1, MARCO, AMN,		
			CD44, CD5		
Small malagula			HBB, HBD,ARG2,		
Sinan molecule	12	1.00E-05 to 4.95E-02	HBA1/HBA2,CYP4B1, MX1, CYTL1,		
biochemistry			CYP4A11, MARCO, MSMO1, DSP		
Amino acid metabolism	4	3.80E-04 to 3.48E-02	ARG2, EDN1, HAL, FRK		
Embrionic development	4	6.80E-04 to 4.40E-02	HBB, HBD, CYTL1, EDN1		
Carbohydrate metabolism	3	7.54E-04 to 4.96E-02	CD244, EDN, CYTL1		
Linid matcheliem	7	7.54 ± 0.4 to 4.04 ± 0.02	CD244, EDN1,CYP4A11, HBB,		
Lipid metabolism	1	7.54E-04 10 4.90E-02	MARCO, MSMO1, DSP		
Reproductive system	2	1.05E.02 to 4.06E.02	NOO1 TNC		
development and function	2	1.95E-05 10 4.90E-02	NQOI, INC		
Protein synthesis	3	1.03E-02 to 2.70E-02	HBA1/HBA2, HBB, ADAMTS4		
Energy production	3	1.64E-03 to 2.43E-02	EDN1, MARCO		
Vitamin and Mineral	2	1 50E 02 to 2 37E 02	EDN1 CD244 CD5		
Metabolism	5	1.50E-02 to 2.57E-02	EDIVI, CD244, CD3		

 Table 9
 Functional categories and corresponding DEGs in high androstenone testis tissues

* Numbers in the *p*-value column showed a range of *p*-values for the genes from each category

IPA assigned 39 of DEGs in high and low androstenone liver group to 6 different canonical pathways. Assigned canonical pathways were metabolic processes including retinol, glycerolipid, fatty acid metabolism and xenobiotics metabolism by Cytochrome P450. Other pathway categories, including PXR/RXR and VDR/RXR activation were also enriched.



Figure 5Functional grouping of DEGs in liver with high and low androstenone using
Ingenuity Pathways Analysis software.

The most significant functional groups (p < 0.05) are presented graphically. The bars represent the *p*-value on a logarithmic scale for each functional group.

ussues			
Function	Number of genes	<i>p</i> -value*	Genes
Amino acid metabolism	3	8.71E-06 to 3.49E-02	HAL, SDS,CDKN1A,
Small Molecule	0	9 71E 06 to 2 51E 02	HAL, CYP7A1, MBL2, AMPD3,
Biochemistry	0	8.71E-00 to 2.51E-02	HSD17B2, IP6K1, SDS, CDKN1A
Cellular Development	4	3.15E-04 to 2.49E-02	CDKN1A, KRT8, HIST1H4A, MBL2
Lipid Metabolism	5	1.10E-03 to 2.41E-02	CYP7A1, MBL2, HSD17B2, IP6K1, CDKN1A, KRT8
Molecular transport	3	1.11E-03 to 4.41E-02	CYP7A1, MBL2, CDKN1A
Cell Function and Maintenance	4	1.20E-03 to 4.90E-02	CDKN1A, MBL2, KRT8, KRT18
Cell Growth and Proliferation	3	1.20E-03 to 2.90E-02	CDKN1A, MBL2, KRT8

Table 10	Functional	categories	and	corresponding	DEGs	in	high	androstenone	liver
	tissues								

* Numbers in the *p*-value column showed a range of *p*-values for the genes from each category

In order to determine the biologically relevant networks other than canonical pathways, network analysis was performed for DEGs in testis and liver samples. The networks describe functional relationships between gene products based on known interactions reported in the literature. Figure 6 exemplarily shows the network deduced from the list of functional candidate genes from testis which are important for androstenone biosynthesis. The network of testis androstenone level comprised of 16 focus genes belonging to functional categories such as molecular transport, haematological disease and haematological system development and function (Figure 6).



Figure 6 Gene network showing the relationship between molecules differentially expressed in high androstenone testis samples.

Genes represented in this network are involved in lipid metabolism, small molecule biochemistry and molecular transport. The network showed a relationship between genes involved in the transport of lipid related molecules (ARL4C and CYP4A11) via blood system (HBA1/HBA2).

The second network of genes from liver androstenone contained 11 focus genes associated with drug metabolism, endocrine system development and function and energy production (Figure 7). The network shows the relationship between beta-estradiol and genes such as FMO5, SMPDL3A and HSD17B2. The gene network shows that retinoid X receptor (RXR) gene had direct relationship between *PLIN2*, *CYP7A1* and *NFkB* genes and indirect relationship with *CDKN1A* gene.



Figure 7 Gene network showing the relationship between molecules differentially expressed in high androstenone liver samples.

Direct or indirect relationships between molecules are indicated by solid or dashed connecting lines, respectively. The type of association between two molecules is represented as a letter on the line that connects them. P, phosphorylation; A, gene activation; E, involved in expression; PP, protein-protein interaction; PD, protein DNA-binding; MB, membership in complex; LO, localization; L, proteolysis; RB, regulation of binding; T, transcription. The number in parenthesis represents the number of bibliographic references currently available in the Ingenuity Pathways Knowledge Base that support each one of the relationships. The intensity of the color in the object is proportional to fold change.

3.4 Validation of selected DEGs with quantitative Real Time PCR (qRT-PCR)

In order to validate the RNA-Seq results, a total of 14 genes were randomly selected and quantified using qRT-PCR. *SULT2A1*, *DHRS4*, *ESR1*, *TNC*, *UCHL1*, *GSTA2* and *CYP2C33* genes from testis samples and *HSD3B1*, *CYP7A1*, *FMO5*, *IGFBP1*, *PLIN2*, *DHRS4* and *HSD17B2* genes from liver samples were selected for the validation by qRT-PCR. Comparison of qRT-PCR data for 14 selected genes showed almost complete concordance of expression with the RNA-Seq results (Figure 8, A and B).



Figure 8qRT-PCR validation for fourteen DEGs from divergent androstenone levels in
(A) testis and (B) liver samples.

Fold change determined via division of high androstenone group gene expression value by low androstenone group gene expression value. Gene expression values for qRT-PCR were normalized using housekeeping genes *PPIA* and *GAPDH*.

3.5 Gene variation analysis

In total 222,225 and 202,249 potential polymorphism were identified in high and low androstenone testis groups. Among these identified polymorphisms, 8,818 in high androstenone group and 8,621 in low androstenone group were global polymorphisms with reference and accession identifiers in dbSNP database. Similarly in liver high and low androstenone samples 169,181 and 164,417 potential polymorphisms were identified. There were 6,851 global polymorphisms in high androstenone liver sample and 6,436 global polymorphisms in low androstenone liver sample.

Polymorphisms identified in DEGs for testis and liver samples are given in Table 11 and Table 12. In the testis samples 12 gene polymorphisms were identified in 8 DEGs (Table 11). Additionally our results revealed that mutations for the genes *CD244* and *ARG2* were specific for high androstenone testis tissues, whereas mutations in genes *IFIT2*, *DSP* and *IRG6* were specific for low androstenone testis samples.

Thirty six mutations were identified in 11 DEGs in liver samples (Table 12). Variation in HAL gene was specific for high androstenone liver samples whereas FMO5, HIST1H4K and TSKU gene variations were specific for low androstenone liver samples (Table 12).

								root mean	
Refseq Id	Gene name	Chr	Position	dbsnp	Reference	Alternate	Quality	square	Sample group
								phred score	
NM_214061	MX1	13	144402807	0	А	G	43.93	37.4	High androstenone
NM_214061	MX1	13	144420441	0	С	Т	59.51	37	Low androstenone
XM_001928671	IFIT2	14	106102335	rs80925743	G	Т	98.81	37	Low androstenone
XM_001928671	IFIT2	14	106102694	0	G	А	40.63	37	Low androstenone
XM_003124689	HBA2	3	35253219	0	G	GA	333.66	39.1	High and Low
XM_003124689	HBA2	3	35253521	0	G	GCTC	617.84	39.89	High and Low
NM_213817	IRG6	3	118838598	0	G	А	181.62	37	Low androstenone
XM_003128168	DSP	7	4940734	0	G	А	31.18	37	Low androstenone
XM_003128168	DSP	7	4944881	0	С	Т	69.47	37	Low androstenone
XM_003124870	LOC100516362	3	48107044	0	G	GT	260.38	38.59	High androstenone
XM_001928325	CD244	4	93149337	0	Т	А	84.61	36.72	High androstenone
XM_001928679	ARG2	7	99786827	0	А	AT	106.28	39.51	High androstenone

Table 11Polymorphisms detected in testis samples

Refseq Id	Gene name	Chr	Position	dbsnp	Reference	Alternate	Quality	root mean	Sample group
NM 214125	MBL2	14	101464163	0	С	Т	236.74	38.58	High and Low
NM 214125	MBL2	14	101464174	0	Т	G	624.35	38.02	High and Low
 NM 214125	MBL2	14	101464216	0	А	Т	2136.83	37.43	High and Low
 NM_214125	MBL2	14	101464268	0	А	G	2123.64	37.42	High and Low
NM 214125	MBL2	14	101464309	0	Т	С	1038.03	37.83	High and Low
 NM_214125	MBL2	14	101464842	0	А	ACT	1693.34	37.6	High and Low
NM_214125	MBL2	14	101467788	0	А	G	4598.53	36.97	High and Low
XM_001928302	SDS	14	38865735	0	G	GT	194.2	39.81	High Androstenone
XM_001928302	SDS	14	38868514	0	С	Т	51.13	37	Low androstenone
XM_001928594	FMO5	4	104473018	rs80837900	G	А	112.94	37.02	Low androstenone
NM_001005352	CYP7A1	4	77195279	0	G	А	1373.52	37.15	Low androstenone
NM_001005352	CYP7A1	4	77195397	0	Т	С	4026.2	36.93	High and Low
NM_001005352	CYP7A1	4	77197364	0	Т	С	1624.62	37	High and Low
NM_001005352	CYP7A1	4	77199510	0	А	Т	416.26	37	Low androstenone
NM_001005352	CYP7A1	4	77199576	0	Т	G	242.3	36.07	Low androstenone
NM_001005352	CYP7A1	4	77200294	0	G	А	274.6	37	Low androstenone
NM_001005352	CYP7A1	4	77200408	rs80923210	А	G	494.32	36.44	Low androstenone
NM_001005352	CYP7A1	4	77201533	rs196960866	А	G	1880.98	37	Low androstenone
NM_001159615	KRT8	5	16715238	rs80814240	С	Т	2365.07	36.65	High and Low
NM_001159615	KRT8	5	16718099	rs80845521	С	Т	2186.76	37.01	High and Low
NM_001159615	KRT8	5	16720725	0	G	GT	1160.84	39.98	High and Low
NM_001159615	KRT8	5	16721108	0	GGT	G	355.1	38.63	High and Low

Table 12Polymorphisms detected in liver samples

Refsea Id	Gene name	Chr	Position	dbsnp	Reference	Alternate	Quality	root mean	Sample group
Kersey lu	Gene hame	CIII	1 OSITIOII	uosnp	Keletenee	Alternate	Quanty	square	Sumple group
NM_001159615	KRT8	5	16721708	rs55618932	Т	С	2941.94	36.61	High and Low
NM_001159615	KRT8	5	16721831	rs80916149	А	G	9343.88	37.02	High and Low
XM_003126180	KRT18	5	16788495	0	G	А	327.13	37.16	Low androstenone
XM_003126180	KRT18	5	16789240	0	G	GA	304.54	38.77	High and Low
XM_003126180	KRT18	5	16789379	rs81211893	G	А	596.36	36.52	High and Low
XM_003126180	KRT18	5	16789412	rs81211894	А	G	2116.22	36.56	High and Low
XM_003126180	KRT18	5	16789808	rs81211895	G	С	2188.63	36.89	High and Low
XM_003126180	KRT18	5	16789954	0	А	ATC	926.56	35.84	High Androstenone
XM_001925061	HAL	5	82556747	0	G	GT	80.57	40.2	High Androstenone
XM_001928022	HIST1H4K	7	22186329	0	С	Т	717.72	33.45	Low androstenone
XM_001929558	CDKN1A	7	36992673	rs80964639	А	G	544.05	37	High Androstenone
XM_001929558	CDKN1A	7	36992792	0	А	G	214.41	37	High and Low
XM_003129674	TSKU	9	10759263	0	G	А	127.72	36.56	Low androstenone
NM_001123146	NNMT	9	40584781	0	G	GA	460.61	42.34	High and Low

3.6 Association between candidate genes and boar taint compounds

The distribution of genotype and allele in all six candidate genes is shown in Table 13. The observed genotype frequencies for FMO1, CYP21, ESR1 and PLIN2 genes differed from those expected from Hardy-Weinberg Law. In case of FMO5 and PLIN22 genes, the observed genotype frequencies were according to expected values from Hardy-Weinberg Law.

Gene	No. of boars	Genotypes			р	q	χ²	p-value
		AA	AC	CC				
FMO1	302	0.03	0.16	0.81	0.11	0.89	13.6	0.0001
		AA	AG	GG				
FMO5	300	0.24	0.52	0.24	0.5	0.5	0.50	n.s.
	-	TT	TC	CC	_			
CYP21	311	0.59	0.2	0.21	0.69	0.31	87.9	0.0001
	-	CC	СТ	TT	_			
ESR1	301	0.84	0.05	0.11	0.865	0.135	187.5	0.0001
		AA	AG	GG				
PLIN2	317	0.76	0.2	0.04	0.86	0.14	7.96	0.05
		AA	AG	GG	_			
PLIN22	340	0.59	0.35	0.06	0.765	0.235	0.70	n.s.

Table 13Genotype frequencies for tested genes.

n.s=non-significant, χ^2 =Chi-square test, *p*-value=deviation from Hardy-Weinberg Law

This study could not observe association of FMO1, PLIN2 and PLIN22 genotypes with boar taint compounds in the population (Table 14). The result of association analysis of FMO5 gene revealed significant association of additive effect and dominance effect with androstenone and skatole respectively. The association analysis result of CYP21 revealed that there were additive effects which involved with varying levels of skatole and indole respectively.

The association analysis of FMO5 gene revealed that animals with homozygote genotype "GG" (6.07) had significantly increased androstenone level, whereas animals with heterozygote genotype "AG" (4.88 and 3.93, respectively) had significantly increased skatole and indole level .The association analysis of CYP21 gene revealed that animals with homozygote genotype "CC" (5.13 and 4.27, respectively) had significantly increased skatole and indole levels whereas in case of ESR1, the result of association analysis revealed that animals with homozygote genotype "TT" and heterozygote genotype "CT" (6.15 and 4.36, respectively) had significantly increased level of androstenone and indole respectively.

Polymorphism	Boar taint compound	Genotype ($\mu \pm S.E.$)			Effect ($\mu \pm S.E.$)	
		AA	AC	CC	Additive	Dominance
FMO1 g.256 A>C	Log Androstenone	5.90±0.31	5.88±0.14	5.94±0.07	-0.02±0.16	0.03±0.21
	Log Skatole	4.40±0.32	4.72±0.14	4.76±0.07	-0.17±0.16	-0.13±0.21
	Log Indole	3.47±0.27	3.80±0.12	3.91±0.06	-0.22±0.14	-0.11±0.18
		AA	AG	GG		
FMO5 g.494 A>G	Log Androstenone	5.70±0.11 ^a	5.96±0.92 ^{ab}	6.07±0.12 °	-0.18±0.08*	-0.07±0.12
	Log Skatole	4.47±0.12 ^a	4.88±0.93 ^b	4.73±0.12 ^{ab}	-0.13±0.08	-0.27±0.12*
	Log Indole	3.64±0.10 ^a	3.93±0.08 ^b	3.90±0.10 ^{ab}	-0.13±0.07	-0.16±0.10
		TT	ТС	CC		
CYP21 g.3911 T>C	Log Androstenone	5.83±0.08	6.00±0.13	6.02±0.16	-0.09±0.08	-0.08±0.70
	Log Skatole	4.63±0.08 ^a	4.78±0.13 ^{ab}	5.13±0.16 [°]	-0.25±0.08*	-0.10±0.17
	Log Indole	3.80±0.07 ^a	3.78±0.11 ^{ab}	4.27±0.13 ^c	-0.24±0.07**	0.26±0.15
		CC	СТ	TT		
ESR1 g.672 C>T	Log Androstenone	5.72±0.07 ^a	5.91±0.25 ^{ab}	6.15±0.16 ^c	-0.21±0.09*	0.02±0.27
	Log Skatole	4.67±0.07	5.23±0.26	4.75±0.16	-0.04±0.09	-0.52±0.27*
	Log Indole	3.78±0.06 ^a	4.36±0.22 ^b	3.73±0.14 ^a	0.02±0.08	-0.60±0.23**

Table 14Genotype and association analysis of candidate genes and boar taint compounds.

Polymorphism	Boar taint compound	Genotype ($\mu \pm S$.)	E.)		Effect ($\mu \pm S.E.$)	
		AA	AG	GG		
PLIN2 g.98 A>G	Log Androstenone	5.92±0.06	6.10±0.14	5.77±0.27	0.07±0.14	-0.25±0.20
	Log Skatole	4.76±0.06	4.80±0.14	4.56±0.28	0.10±0.15	-0.14±0.20
	Log Indole	3.85±0.05	3.92±0.12	3.90±0.24	-0.02±0.12	-0.03±0.18
		AA	AG	GG		
PLIN22 g.198 A>G	Log Androstenone	5.78±0.07	6.06±0.10	6.00±0.20	-0.11±0.11	-0.17±0.15
	Log Skatole	4.67±0.07	4.87±0.10	4.57±0.21	0.05±0.11	-0.25±0.15
	Log Indole	3.78±0.06	3.98±0.09	3.89±0.18	-0.05±0.09	-0.14±0.13

a, b,c * : *P* < 0.05, d,e,f ** : *P* < 0.001, Ln=natural log

4 Discussion

This study showed whole genome expression differences for varying androstenone levels in testis and liver tissues. RNA-Seq provided high resolution map of transcriptional activities and genetic polymorphisms in these tissues. However, due to incomplete porcine annotations, only around 50% of the total reads could be mapped to annotated references. The improvements in pig genome annotations may lead to better coverage and detailed understanding of genetic and functional variants such as novel transcripts, isoforms, sequence polymorphisms and non-coding RNAs. Integration of high throughput genomic and genetic data (eQTL) with proteomic and metabolomic data can provide additional new insight into common biological processes and interaction networks responsible for boar taint related traits.

On the basis of number of DEGs, our results confirm that transcriptome activity in testis is higher in comparison to liver tissue for androstenone biosynthesis. These results also show that the entire functional pathway involved in androstenone metabolism is not completely understood and through this study, we propose additional functional candidate genes such as *SLC22A20*, *DKK2* and *AMN* in testis and *HAMP*, *LOC100512122* and *AADAT* in liver. Furthermore, various gene polymorphisms were also detected in testis and liver DEGs. Potential polymorphisms were identified in DEGs such as *HSP40*, *RASL11A* and *PDZK11P1* in testis and *PLIN2*, *IGFBP1*, *CYP7A1* and *FMO5* in liver. These polymorphisms may have an impact on the gene activity ultimately leading to androstenone variation and could be used as biomarkers for boar taint related traits. Additionally, these potential biomarkers can also be targeted for fertility and reproduction traits while breeding for boar taint. However, further validation is required to confirm the effect of these biomarkers in other animal populations.

Furthermore this study revealed some significant results regarding the reduction of boar taint and enhancing the fertility of boars which is the key question raised by animal breeders and economists. It is not only important to cope up with problem of boar taint but this is equally important that genes treating with boar taint should not affect the reproduction in boars. Gunawan et al. (2011) reported the association of similar SNP of ESR1 with high sperm quality and fertility traits. This aspect revealed the significance of this SNP as far as boar taint and fertility in boars is concerned.

5 Summary

Boar taint is an unpleasant smell and taste of pork meat derived from some entire male pigs. The main causes of boar taint are the two compounds androstenone (5α -androst-16-en-3-one) and skatole (3-methylindole). It is crucial to understand the genetic mechanism of boar taint to select pigs for lower androstenone levels and thus reduce boar taint. The aim of the present study was to investigate transcriptome differences in boar testis and liver tissues with divergent androstenone levels using microarrays and RNA deep sequencing (RNA-Seq).

The total number of reads produced for each testis and liver sample ranged from 13,221,550 to 33,206,723 and 12,755,487 to 46,050,468, respectively. In testis samples 46 genes were differentially regulated whereas 25 genes showed differential expression in the liver. The fold change values ranged from -4.68 to 2.90 in testis samples and -2.86 to 3.89 in liver samples. Differentially regulated genes in high androstenone testis and liver samples were involved in metabolic processes such as lipid metabolism, small molecule biochemistry and molecular transport.

This study provides evidence for transcriptome profile and gene polymorphisms of boars with divergent androstenone level using RNA-Seq technology. Digital gene expression analysis identified candidate genes in flavin monooxygenease family, cytochrome P450 family and hydroxysteroid dehydrogenase family. Moreover, gene polymorphism analysis revealed potential mutations in *IRG2*, *DSP*, *IFIT2*, *CYP7A1*, *FMO5* and *CDKN1A* genes in both high and low androstenone sample groups. Further studies are required for proving the role of candidate genes to be used in genomic selection against boar taint in pig breeding programs.

Additionally six genes FMO1, FMO5, CYP21, ESR1, PLIN2 and PLIN22 were selected for association analysis based on their known function and their differential expression for boar taint compounds. For the association studies, the SNP of six genes were genotyped in a total of 370 animals. Three genes (FMO5, CYP21 and ESR1) were associated with boar taint compounds. In detail, the association analysis of FMO5 showed its significant association with all three boar taint compounds i.e., androstenone, skatole and indole whereas, ESR1 association results showed the association with androstenone and indole. According to the results of association studies, FMO5, CYP21 and ESR1 turned out to be the most promising candidates for boar taint.

6 Zusammenfassung

Ebergeruch ist eine unangenehme Geruchs- und Geschmacksabweichung im Schweinefleisch von Ebern. Ebergeruch wird hauptsächlich durch die Stoffe Androstenon (5α-androst-16-en-3-one) and Skatol (3-methylindole) hervorgerufen. Für die Selektion von Schweinen bezüglich eines geringeren Androstenon- und Skatolgehalts, sowie einer damit verbundenen geringeren Häufigkeit von Geruchsabweichungen, ist es notwendig, die grundlegenden genetischen Mechanismen zu identifizieren. Das Ziel dieser Studie war es, Transkriptom-Differenzen im Testis- und Leber-Gewebe von Tieren mit einem unterschiedlichen Androstenon-Gehalten anhand von Microarray-Chips und der RNA-Sequenzierung (RNA-Seq) zu untersuchen. Insgesamt 13,221,550 und 33,206,723 Sequenzen wurden für die Testis-Proben generiert sowie 12,755,487 und 46,050,468 für die Leber-Proben.

Differentiell reguliert waren im Testis-Gewebe 46 Gene und im Leber-Gewebe 25 Gene. Die "fold change"-Werte variierten zwischen -4.68 und 2.90 in den Testis-Proben und zwischen -2.86 to 3.89 in den Leber-Proben. Die differentiell regulierten Gene aus der "Hoch-Androstenon-Gruppe" waren an den metabolischen Prozessen Fettstoffwechsel, Biochemie kleiner Moleküle und molekularer Transport beteiligt. Anhand der RNA-Sequenzierung wurden in dieser Studie Transkriptom-Profile und Polymorphismen von Ebern mit deutlich unterschiedlichen Androstenon-Gehalten dargestellt. Die Genexpressionsanalyse identifizierte die Kandidatengene in den flavin monooxygenease, cytochrome P450 und hydroxysteroid dehydrogenase Genfamilien. Weiterhin identifizierte die Polymorphismus-Analyse Mutationen in den Genen *IRG2, DSP, IFIT2, CYP7A1, FMO5* und *CDKN1A* sowohl in der hohen als auch in der niedrigen Androstenon Gruppe. Weitere Studien sind notwendig, um die Bedeutung der Kandidaten-Gene zu analysieren, bevor diese für die Genomische Selektion gegen Ebergeruch in Zuchtprogrammen genutzt werden können.

Auf Grund ihrer Funktion und ihrer differentiellen Expression wurden die Gene FMO1, FMO5, CYP21, ESR1, PLIN2 and PLIN22 für Assoziations-Studien ausgewählt. 370 Tiere wurden für SNPs dieser Gene genotypisiert. Die Gene *FMO5*, *CYP21* und *ESR1* zeigten Assoziationen zu den Ebergeruchs Merkmalen, wobei *FMO5* signifikante Assoziationen zu Androstenon, Skatol und Indol zeigte. *ESR1* war mit Androstenon und Indol assoziiert. Die Assoziationsstudie zeigte, dass *FMO5*, *CYP21 and ESR1* vielversprechende Kandidatengene für Ebergeruchsmerkmale sind.

7 Consequences for practical agriculture

It is obvious that castration of piglets with anesthesia will only be accepted as a transitional step until castration will be completely banned in Europe. However, if intact boars are fattened, negative consumer response to boar taint in pork has to be prevented: by testing carcasses routinely with sufficient speed and accuracy and by reducing the incidence of boar taint at slaughter age. This may be approached in different ways: by genetic selection, nutrition and/or management.

On first sight, genomic selection may seem to offer a quick and easy solution. Before drawing premature conclusions, the results of Grindflek et al. (2010) should be noted who found markers for fertility traits on the same locations of the chromosome as for androstenone level, which is not surprising in view of the described antagonistic effects. Moreover associations between markers and traits are known to be breed specific. In any case, genetic markers have to be identified in each population, with relevant correlations to other traits, before genomic selection is applied in practice.

The intensity of boar taint in carcasses of intact boars can be reduced by selection. This can help the pork industry in gradually reducing the number of carcasses discarded because of boar taint and eventually eliminate the need for castration. To achieve optimal response to selection, standardized procedures for measuring the two main components of boar taint, androstenone and skatole, should be developed. Two current research projects (Anon., 2009a,b) are focused on the development of automated measurement of boar taint for use in processing plants as well as on live animals. The eventual goal is to develop techniques for screening live boars for taint score, based on microbiopsy of backfat, saliva or blood samples, which would speed up genetic progress.

The rate at which genetic progress can be reached will depend on antagonistic correlation between boar taint and reproductive traits. These genetic correlations have to be determined in relevant commercial male and female lines.

When identified QTLs for boar taint are being used in genomic selection, special attention should be on gene locations which are not known to be negatively correlated with reproductive performance.

8 Schlussfolgerungen für die Umsetzung der Ergebnisse in die Praxis

Grundsätzlich lässt sich der Anteil genussuntauglicher Eberschlachtkörper züchterisch reduzieren. Voraussetzung hierfür ist jedoch, dass die Erfassung der beiden Leitkomponenten Skatol und Androstenon standardisiert ist und damit eine laborübergreifende Vergleichbarkeit ermöglicht wird. Derzeit werden im Rahmen von zwei Forschungsprojekten (Anon, 2009a,b) die Möglichkeiten einer automatisierten Erfassung des Ebergeruchs für züchterische Zwecke und zur Sortierung im Schlachtprozess untersucht. Die Entwicklung von Technologien zur routinemäßigen Erfassung des Ebergeruchs am lebenden Zuchteber mit Hilfe von Mikrobiopsie-, Speichel- oder Blutproben wären im Sinne schneller Zuchterfolge anzustreben.

Der Erfolg entsprechender Zuchtprogramme wird in entscheidender Weise durch das Ausmaß der zu erwartenden antagonistischen Beziehungen zwischen Reproduktionsmerkmalen und Ebergeruch beeinflusst. Entsprechende populationsspezifische Untersuchungen sollten durchgeführt werden, um die Vereinbarkeit beider Selektionsziele beurteilen zu können.

Durch die Berücksichtigung identifizierter QTL im Rahmen der Genomischen Selektion ist eine Steigerung der Selektionserfolge zu erwarten. Besonderes Augenmerk ist dabei auf Genorte zu legen, mit deren Hilfe die gegenläufige Beziehung der beiden Merkmalskomplexe Fruchtbarkeit und Ebergeruch aufgebrochen werden kann.

9 Consequences for further research

Results concerning the functional pathway involved in androstenone and skatole metabolism will be integrated into the project STRAT-E-GER, Strategien zur Vermeidung von Geruchsabweichungen bei der Mast unkastrierter männlicher Schweine (Fattening entire male pigs - Strategies to prevent boar taint compounds), funded by the Bundesministerium für Ernährung Landwirtschaft und Verbraucherschutz (BMELV), within the programme Innovationsförderung. Association studies may confirm the biological significance of the relevant genes.

10 Patents

11 Publications

- Neuhoff C, Pröll M, C. Große-Brinkhaus, L. Frieden, A. Becker, A. Zimmer, M.U. Cinar, E. Tholen, C. Looft, K. Schellander (2011): Identifizierung von relevanten Genen des Metabolismus von Androstenon und Skatol in der Leber von Jungebern mit Hilfe von Transkriptionsanalysen. Vortragstagung der Deutschen Gesellschaft für Züchtungskunde e.V. (DGfZ) und der Gesellschaft für Tierzuchtwissenschaften e.V. (GfT), 6/7.9.2011, Freising-Weihenstephan, Deutschland
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- Gunawan, A., Sahadevan S., Neuhoff, C., Große-Brinkhaus, C., Tesfaye, D., Tholen, E. Looft, C., Schellander, K., Cinar, M.U. (2012): Using RNA-Seq for transcriptome profiling in liver of boar with divergent skatole levels, P2035, ISAG meeting, Cairns, Australien, 15.7.-20.7.2012
- Neuhoff, C., Pröll, M., Große-Brinkhaus, C., Frieden, L., Becker, A., Zimmer, A., Tholen, E., Looft, C., Schellander, K. and Cinar, M.U. (2102): Global gene expression analysis of liver for androstenone and skatole production in the young boars. p. 274, EAAP meeting, Bratislava, Slovakia, 27.8.-31.8.2012

- Gunawan, A., Sahadevan S., Neuhoff, C., Große-Brinkhaus, C., Tesfaye, D., Tholen, E. Looft, C., Schellander, K., Cinar, M.U. (2012): RNA deep sequencing analysis for divergent androstenone levels in Duroc × F2 boars. Vortragstagung der Deutschen Gesellschaft für Züchtungskunde e.V. (DGfZ) und der Gesellschaft für Tierzuchtwissenschaften e.V. (GfT), 12/13.9.2012, Halle a.d. Saale, Deutschland
- Gunawan, A., Sahadevan S., Neuhoff, C., Große-Brinkhaus, C., Tesfaye, D., Tholen, E. Looft, C., Schellander, K., Cinar, M.U. (2012): RNA deep sequencing reveals novel candidate genes and polymorphisms in boar testis and liver tissues with divergent androstenone levels, BMC Genomics, submitted

12 Presentations

- Neuhoff C. (2011): Identifizierung von relevanten Genen des Metabolismus von Androstenon und Skatol in der Leber von Jungebern mit Hilfe von Transkriptionsanalysen.
 Vortragstagung der Deutschen Gesellschaft für Züchtungskunde e.V. (DGfZ) und der Gesellschaft für Tierzuchtwissenschaften e.V. (GfT), 6/7.9.2011, Freising-Weihenstephan, Deutschland
- Neuhoff, C. (2012): Global gene expression analysis of liver for androstenone and skatole production in the young boars. p. 274, EAAP meeting, Bratislava, Slovakia, 27.8.-31.8.2012
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13 Abstract

Boar taint is an unpleasant smell and taste of pork meat derived from some entire male pigs. The main causes of boar taint are the two compounds androstenone (5α -androst-16-en-3-one) and skatole (3-methylindole). It is crucial to understand the genetic mechanism of boar taint to select pigs for lower androstenone levels and thus reduce boar taint. The aim of this study was the identification of genes and pathways influencing boar taint and involved in androstenone and skatol metabolism. Therefore polymorphisms in relevant genes were identified and transcriptome analysis using Affymetrix-Chips and RNA-Seq in the two major organs involved in androstenone metabolism i.e the testis and the liver was performed.

Differentially regulated genes in high androstenone testis and liver samples were involved in metabolic processes such as retinol metabolism, metabolism of xenobiotics by cytochrome P450 and fatty acid metabolism. Moreover, a number of genes encoding biosynthesis of steroids were highly expressed in high androstenone testis samples. Gene polymorphism analysis revealed potential mutations in *HSP40*, *IGFBP1*, *CYP7A1* and *FMO5* genes affecting androstenone levels. Further studies are required for verify the role of candidate genes to be used in genomic selection against boar taint in pig breeding programs. According to the results of association studies, FMO5, CYP21 and ESR1 turned out to be the most promising candidates for boar taint.

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