Variability in porcine microRNA genes and its association with mRNA expression and lipid phenotypes

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Abstract

Background

Mature microRNAs (miRNAs) play an important role in repressing the expression of a wide range of mRNAs. The variability of miRNA genes and their corresponding 3'UTR binding sites might disrupt canonical conserved miRNA-mRNA pairing, thus modifying gene expression patterns. The presence of polymorphic sites in miRNA genes and their association with gene expression phenotypes and complex traits has been poorly characterized in pigs so far.

Results

By analyzing whole-genome sequences from 120 pigs and wild boars from Europe and Asia, we have identified 285 single nucleotide polymorphisms (SNPs) mapping to miRNA loci, as well as 109,724 SNPs located in predicted 7mer-m8 miRNA binding sites within porcine 3'UTRs. Porcine miRNA genes show a reduced SNP density compared with their flanking non-miRNA regions. By sequencing the genomes of 5 Duroc boars, we have identified 12 miRNA SNPs that have been subsequently genotyped in their offspring (N = 345, Lipgen population). Association analyses between miRNA SNPs and 38 lipid-related traits as well as hepatic and muscle microarray expression phenotypes recorded in the Lipgen population were carried out. The most relevant association detected was the one between the genotype of the rs319154814 (G/A) SNP located in the apical loop of the ssc-miR-326 hairpin precursor and *PPP1CC* mRNA levels in the liver (*q*-value = 0.058). This result was subsequently confirmed by qPCR (*P*-value = 0.027). The rs319154814 (G/A) genotype was also associated with several fatty acid composition traits.

Conclusions

Porcine miRNA genes show a reduced variability consistent with strong purifying

selection, particularly in the seed region, which plays a critical role in miRNA binding.

Although it is generally assumed that SNPs mapping to the seed region are the ones

with the strongest consequences on mRNA expression, we show that a SNP mapping to

the apical region of ssc-miR-326 is significantly associated with the hepatic mRNA

levels of the *PPP1CC* gene, one of its predicted targets. Although experimental

confirmation of such interaction has been obtained in humans but not in pigs, this result

highlights the need of further investigating the functional effects of miRNA

polymorphisms located outside the seed region on gene expression in pigs.

Background

Mature microRNA transcripts (miRNAs) are short (~22 nt) non-coding RNAs

which play an essential role in the regulation of gene expression [1]. During the

biogenesis of miRNAs, one strand of the miRNA duplex binds to the guide-strand

channel of an Argonaute protein forming a miRNA-induced silencing complex

(miRISC) with the ability to repress mRNA expression through binding to specific

3'UTR target sites [2]. In postembryonic cells, this repressor mechanism mainly acts by

destabilizing the mRNA through decapping and poly(A)-tail shortening [3,4] and less

often by hindering translation [2]. The binding of the miRNA to its 3'UTR target site

depends critically on the sequence of the seed region, which encompasses nucleotides

(nt) 2-8 from the 5' end of the mature miRNA and interacts with the target site through

Watson-Crick pairing [1]. Polymorphisms located within the seed region might affect

the proper recognition of mRNA targets [5,6]. Nevertheless, imperfect seed matches can

be compensated by nucleotides 13th-to-16th of the mature miRNA, thus providing

additional anchor pairing to the seed region [2]. Other sites relevant for miRNA

processing and function are the basal UG, flanking CNNC and apical UGU motifs [7],

as well as a mismatched GHG motif [8], all of which contribute to facilitate miRNA

processing [2]. Variability in the 3'UTR miRNA binding sites can also affect gene

expression at the post-transcriptional level [5,9,10].

The evolutionary conservation of miRNAs across species is associated with their

levels of expression [11,12] as well as with the functional importance of the regulatory

networks they modulate [13,14]. Besides, the conservation of the repertoire of mRNAs

targeted by a given miRNA depends mostly on the age of the associated miRNA gene,

with novel miRNAs acquiring targets more rapidly than ancient miRNAs [15].

Conserved sites are particularly enriched in the 5'end of 3'UTR [12], although selective

pressure on miRNA binding sites is not uniform across miRNA genes [16].

Saunders et al. (2007) [5] investigated the variability of 474 human miRNA

genes and found that the single nucleotide polymorphism (SNP) density is lower in

miRNA loci when compared to their flanking regions. Moreover, Saunders et al. (2007)

[5] found that ~90% of human miRNA genes do not contain polymorphisms, and the

majority of SNPs mapping to miRNAs are located outside the seed region, thus

evidencing that the variability of this critical functional motif evolves under strong

selective constraints [5]. Indeed, polymorphisms in the first 14 nucleotides of the

mature miRNA, and particularly those within the seed region, might abrogate the

binding of the miRNA to its 3'UTR targets, leading to an extensive rewiring of the

miRNA-mediated regulatory network [17]. Mammalian miRNA knockouts often

display abnormal phenotypes, reduced viability and clinical disorders [2], although

functional redundancy amongst miRNA family members might mitigate to some extent

the severity of such manifestations [2]. Polymorphisms within miRNA loci laying

outside the seed region can also affect the processing and stability of miRNAs during

their maturation and loading into the functional silencing complex [18,19]. Moreover,

Saunders et al. (2007) [5] showed that a broad array of predicted miRNA target sites in

the 3'UTR of mRNAs are polymorphic, a finding that suggests that purifying selection

on these regions is less intense than in miRNA genes [17]. Overall, these findings

suggest functional conserved mechanisms of miRNA-mediated gene regulation

influenced by polymorphic sites at both miRNA loci and their 3'UTR binding sites [20].

The wild ancestors of pigs were independently domesticated in the Near East

and China 10,000 YBP [21]. Domestic pigs then spread worldwide, becoming one of

the most important sources of animal protein for humans and diversifying in an

extensive array of breeds with distinct morphological and productive features [22]. Pig

phenotypes might be explained, at least partially, by modifications in the microRNA-

mediated regulation of gene expression [23]. Indeed, several studies have reported

associations between SNPs in miRNA binding sites and porcine phenotypic variation

[24–29], while fewer studies have investigated the association between SNPs in miRNA

genes and complex traits in pigs [30–33]. For instance, Chai et al. (2018) reported a

significant causal relationship between a SNP in the seed of ssc-miR-378-3p and

extensive modifications of its target mRNA repertoire [33]. Moreover, the authors

associated the presence of the observed miRNA seed variant to a structural

reconfiguration of the hairpin, leading to an enhanced expression of the mature miRNA

[33].

The polymorphism of miRNA genes has not been systematically characterized

in pigs despite its potential impact on gene regulation and phenotypic variation. In the

current work, we aimed to elucidate the patterns of variability of miRNA genes in European and Asian wild boars and domestic pigs as well as to infer whether such patterns are influenced by purifying selection. Moreover, we wanted to infer whether SNPs in miRNA genes are associated with liver and muscle gene expression traits and lipid phenotypes recorded in a Duroc pig population.

Methods

Characterizing the polymorphisms of miRNA genes and their 3'UTR binding sites in a worldwide sample of pigs and wild boars

Retrieval of Porcine Whole-Genome Sequences

Whole-genome sequences (WGS) from 120 wild boars and domestic pigs (*Sus scrofa*) were retrieved from the NCBI Sequence Read Archive (SRA) database (https://www.ncbi.nlm.nih.gov/sra). Most of these WGS have been reported in previous publications [34–37] and detailed information can be found in **Additional file 1: Table S1**. The 120 WGS corresponded to Asian domestic pigs (ADM, N = 40), Asian wild boars (AWB, N = 20), European domestic pigs (EDM, N = 40) and European wild boars (EWB, N = 20). The ADM group included WGS from Meishan, Tongchen, Jinhua, Rongchan, Wuzhishan, Tibetan, Sichuan, Hetao, Minzhu, Bamaixang and Laiwu pigs, while the EDM group was composed of WGS from Pietrain, Mangalitza, Iberian, Duroc, American Yucatan (from America, but with a European origin), Yorkshire, Landrace, Hampshire and Large-white pigs. Both AWB and EWB samples were selected according to their geographic distribution (**Additional file 1: Table S1**).

The EWB group contained one sample from the Near East. Raw data in SRA format

were downloaded from SRA public repositories and converted into fastq format by

using the fastq-dump 2.8.2 tool available in the SRA-toolkit package

(ncbi.github.io/sra-tools/).

Whole-genome sequence data processing and calling of single nucleotide

polymorphisms

Fastq paired-end files generated from SRA data were filtered according to their

quality, and sequence adapters were trimmed by making use of the Trimmomatic

software (v.0.36) with default parameters [38]. Trimmed paired-end sequences were

aligned against the *Sus scrofa* reference genome (Sscrofa11.1) [39] with the BWA-

MEM algorithm [40] and default settings. Sequence alignment map (SAM) formatted

files were sorted and transformed into binary (BAM) formatted files. PCR duplicates

were subsequently identified with the *MarkDuplicates* package from the Picard tool

(https://broadinstitute.github.io/picard/) and removed to perform INDEL realignment

with the *IndelRealigner* package from the Genome Analysis Toolkit (GATK v.3.8) [41].

Base quality score recalibration (BQSR) was implemented with the *BaseRecalibrator*

package from GATK v.3.8. Variant calling was implemented with the *HaplotypeCaller*

function according to GATK best practices [41]. Individual gVCF formatted files,

including both polymorphic and homozygous blocks, were generated with the

GenotypeGVCFs package from the GATK v.3.8 tool [41], and they were merged into

separate multi-individual variant calling format (VCF) files containing single

polymorphic and INDEL sites for each defined population (i.e. ADM, EDM, AWB and

EWB). Finally, variants were distributed into SNP and INDEL files with the

SelectVariants package and quality filtered using the *FilterVariants* function from

GATK v.3.8 [41]. For SNPs, we used the following parameters: QD < 2.0, FS > 60.0,

MQ < 40.0, MQRankSum < -12.5, ReadPosRankSum < -8.0, SOR > 3.0. INDELs were

not taken into consideration in the present study because INDEL detection based on

commonly used pipelines for variant calling have shown low reliability and

reproducibility across tools and analyzed datasets [42,43].

Identification of single nucleotide polymorphisms in miRNA genes and in their 3'UTR

binding sites

Single nucleotide polymorphisms mapping to annotated porcine miRNA loci (N

= 370) were retrieved by making use of the curated Sscrofa11.1 annotation for miRNA

regions available in the miRCarta v1.1 [44] and miRBase [45] databases. Additionally,

annotated mature miRNA loci (N = 409) within miRNA genes (N = 370) were

retrieved, and SNPs located in the mature and seed regions (2nd to 8th positions at 5' end

in the mature miRNA) were identified. A comprehensive description of miRNA gene

regions and miRNA-mRNA interactions can be found in **Fig. 1A** and **1B**, respectively.

We also retrieved the annotated 3'UTR porcine mRNA transcripts from

Ensembl repositories (v.100, http://www.ensembl.org/info/data/ftp/index.html) and the

corresponding set of sequences was interrogated against the seed regions of the 409

porcine mature miRNAs considered in our study. Seed sequences were reverse-

complemented and searched along the 3'UTR sequences of mRNA genes by making

use of the *locate* function of the SeqKit toolkit [46]. All 7mer-m8 canonical seed pairing

sites, as well as 8mer seed pairing sites (**Fig. 1**), were identified for each miRNA seed,

and, based on such information, the corresponding putative miRNA-mRNA target pairs

were established. Additional non-canonical miRNA-mRNA interactions were not

considered (**Fig. 1B**) as they are regarded as less biologically significant than 7mer-m8

or 8mer interactions [2,47,48]. Subsequently, the genomic location of both 7mer-m8

and 8mer matching regions were determined, and SNPs residing in such predicted

miRNA binding sites were gathered.

Principal components analyses (PCA) were performed with the smartPCA

software [49]. These PCA analyses were based on either (1) autosomal whole-genome

SNPs, (2) autosomal SNPs from miRNA genes, (3) autosomal SNPs from 3'UTRs, or

(4) autosomal SNPs from 8mer and 7mer-m8 3'UTR sites. When implementing the

PCA based on whole-genome data, SNPs complying with the following parameters

were retained: 1) minimum allele frequency > 0.05, and 2) Hardy-Weinberg equilibrium

exact test *P*-value > 0.001. In contrast, no filtering was performed in the PCA based on

miRNA SNPs because there are a few hundreds.

Frequency and distribution of SNPs in miRNA genes

To investigate the distribution of SNPs along miRNA genes, SNPs were

classified according to their location, i.e., SNPs within the seed region of mature

miRNAs were flagged as "*seed*", whereas SNPs located in mature miRNAs but outside

the seed were classified as "*mature*". Additionally, SNPs classified as "*mature*" were

assigned to the following subtypes: 1) "*anchor*" (1st position at 5' end), and 2)

"*supplemental pairing*" (13th to 18th position from 5' end). The remaining SNPs were

assigned to the "*precursor*" class. Allele frequencies were estimated in the whole set of

porcine samples (N = 120), as well as in each EDM, ADM, EWB and AWB group

independently (**Additional file 2: Table S2**).

In order to calculate the SNP density in precursor, mature and seed miRNA

regions, we first calculated the total nucleotide length of each of these regions in the set

of porcine miRNA loci. S*eed length* was calculated by considering 7 nucleotide

positions in each of the annotated mature miRNAs (N = 409) mapping to the 370

miRNA loci considered in our study (7 bp × 409 = 2,863 bp). The total length of the

409 mature miRNAs was determined (~22 bp × 409 = 8,916 bp), and the total *seed*

length (2,863 bp) was subtracted from 8,916 bp to obtain the *mature length* (6,053 bp),

which corresponds to the total length of all mature miRNA sequences excluding the

seeds. The *precursor length* (22,229 bp) was calculated by subtracting the *seed length*

(2,863 bp) and *mature length* (6,053 bp) from the total length (31,145 bp) of all miRNA

loci (N = 370). Then, the SNP density (*D*) in each of these regions was calculated as

follows:

$$D = \frac{N_r \times 100}{L_r}$$

where N_r is the total number of miRNA SNPs (N = 285), or the number of SNPs

detected in each of the defined regions, (i.e., 221, 52 and 12 SNPs in precursor, mature

and seed regions, respectively); L_r is the total nucleotide length of all miRNA loci

(31,145 nts), or the total number of nucleotides in miRNA loci belonging to the

"*precursor*" (22,229 bp), "*mature*" (6,053 bp) or "*seed*" (2,863 bp) regions. These

calculations would yield the number of SNPs per bp for each category. We decided to

adjust this estimate to a window of 100 bp, which results in the multiplication of N_r by

100 in the above formula. We also compared the SNP density between miRNA loci (N

= 370) and their flanking genomic regions (± 1 kb) by considering 10 upstream and

downstream nucleotide bins comprising 100 bp each. Differences in SNP density across

miRNA loci, as well as compared with surrounding regions, were estimated with a non-

parametric approach using a Mann Whitney U test [50].

Investigating the association of miRNA polymorphisms with gene expression and

phenotype data recorded in Duroc pigs

Whole-Genome sequencing of five Duroc pigs

In the year 2,003, five Duroc boars were selected and used as founders of a half-

sib population of purebred Duroc pigs devoted to the production of high-quality cured

ham. By whole-genome sequencing of these 5 individuals, we aimed to identify miRNA

SNPs and investigate their association with gene expression traits and lipid phenotypes

recorded in the offspring of the 5 boars (Lipgen population, N = 345). Genomic DNA

was extracted [51] and sequenced at the Centre Nacional d'Anàlisi Genòmica (CNAG,

Barcelona, Spain). Paired-end multiplex libraries were prepared according to the

instructions of the manufacturer with the KAPA PE Library Preparation kit (Kapa

Biosystems, Wilmington, MA). Libraries were loaded to Illumina flow-cells for cluster

generation prior to producing 100 bp paired-end reads on a HiSeq2000 instrument

following the Illumina protocol. Base calling and quality control analyses were

performed with the Illumina RTA sequence analysis pipeline according to the

instructions of the manufacturer. Quality-checked filtered reads were mapped to the *Sus*

scrofa genome version 11.1 and processed for SNP calling according to GATK best

practices recommendations [41] and the same protocol defined in a previous section

(please see the section describing Whole-genome sequence data processing and calling

of single nucleotide polymorphisms). Since only five boars were sequenced, SNPs

within miRNA loci were retrieved without quality filtering criteria, in an effort to

maximize the number of identified miRNA SNPs.

Recording of phenotypes related with gene expression and lipid traits in the Lipgen

population

A total of 350 Duroc barrows, sired by the five Duroc founder boars mentioned

before were used as a resource population (Lipgen population [52,53]) to evaluate

associations between miRNA SNPs and phenotypes related with gene expression and

lipid traits. The five sequenced boars were mated with 400 sows in three different farms

and one offspring per litter was selected for phenotypic recording (only 350 individuals

provided valid records). All selected piglets were weaned, castrated and subsequently

fattened in four contemporary batches at the IRTA pig experimental farm in Monells

(Girona, Spain) under intensive standard conditions. Once they reached ~122 kg of live

weight (~190 days of age), they were slaughtered in a commercial abattoir following the

guidelines of current Spanish legislation (https://www.boe.es/buscar/doc.php?id=BOE-

A-1995-3942). After slaughtering, tissue samples from *gluteus medius* (GM) and

longissimus dorsi (LD) skeletal muscles and liver were obtained according to previously

described methods [54–56].

Total DNA was extracted from each sample following Vidal et al. 2005 [51]. A

subset of 345 DNA samples from the initial set of 350 pigs were successfully obtained

and processed for further genotyping. Total RNA was extracted from GM (N = 89 pigs)

and liver (N = 87 pigs) tissue samples by using the Ribopure isolation kit (Ambion,

Austin, TX). Expression mRNA profiles were characterized by hybridizing total RNA

to GeneChip porcine arrays (Affymetrix Inc., Santa Clara, CA), which encompass a

total of 23,998 probes [54,57]. Further details about tissue collection, sample selection,

RNA isolation and microarray hybridization procedures can be found in Cánovas et al.

(2010) [54]. Microarray data pre-processing, background correction, normalization and

log_2-transformation of expression estimates were performed with a robust multi-array

average method per probe [58]. The *mas5calls* function from the affy R package [58]

was then applied to identify probes displaying intensity signals above the background

noise. This function applies a Wilcoxon signed rank-based gene expression

presence/absence detection algorithm for labeling expressed probes in each sample.

Control probes and those with expression levels below the detection threshold in more

than 50% of samples were discarded from further analyses. By doing so, we only

retained probes corresponding to mRNAs that were robustly expressed in GM muscle

(12,131 probes) and liver (12,787 probes) tissues. Probes from GeneChip Porcine

Genomic array identifiers (Affymetrix, Inc., Santa Clara, CA) of GM skeletal muscle

and liver tissues were assigned to mRNA genes annotated in the Sscrofa11.1 assembly

[39] using the BioMart tool [59].

With regard to lipid-related phenotypes, 38 traits were measured in the Duroc

Lipgen population, i.e. backfat thickness measured between 3rd and 4th ribs, backfat

thickness at the last rib, ham fat thickness, and intramuscular fat content and

composition of GM and LD skeletal muscle samples (N = 345) [53,55] (**Additional file**

3: Table S3). Briefly, intramuscular fatty acid (IMF) content in the GM and LD

muscles was measured with the Near Infrared Transmittance technique (NIT, Infratec

1625, Tecator Hoganas, Sweden), while muscle cholesterol measurements were inferred

following Cayuela et al. [60]. Gas chromatography of methyl esters was used to

determine muscle fatty acids (FA) composition, including the percentages of saturated

(SFA), unsaturated (UFA), monounsaturated (MUFA) and polyunsaturated (PUFA)

fatty acids. Live and carcass weights (used as covariates in the statistical analyses), as

well as backfat and ham fat thickness, were measured on a regular basis prior (live

weight and backfat thickness) and after slaughter (carcass weight and ham fat

thickness). Mean and standard deviations of lipid traits recorded the Lipgen population

(N = 345) are described in **Additional file 3: Table S3**.

Genotyping of a panel of single nucleotide polymorphisms mapping to microRNA genes

in the Lipgen population

Twelve SNPs mapping to miRNA loci and segregating in the five sequenced

parental Duroc boars were genotyped in the 345 Duroc pigs from the Lipgen population

(**Table 1**). Briefly, selected miRNA SNPs and their flanking regions (60 upstream and

downstream bp) were used for assay design with the Custom TaqMan Assay Design

Tool   website   (https://www5.appliedbiosystems.com/tools/cadt/;   Life   Technologies).

Genotyping tasks were carried out at the Servei Veterinari de Genètica Molecular of the

Universitat Autònoma of Barcelona (http://sct.uab.cat/svgm/en) by using a QuantStudio

12K Flex Real-Time PCR System (Thermo Fisher Scientific, Barcelona, Spain).

Association analyses between miRNA SNPs and mRNA expression and lipid phenotypes

Genotype data corresponding to the 12 miRNA SNPs mentioned before (**Table**

1) were processed with the PLINK software [61] in order to generate formatted files for

subsequent   analyses.   The   Genome-Wide   Efficient   Mixed-Model   Association

(GEMMA) software [62] was used to implement association analyses between

genotyped SNPs and phenotypes related with microarray gene expression data and lipid

traits. The following univariate mixed model was used:

$$y = W\alpha + x\delta + u + \varepsilon$$

Where y is the phenotypic vector of recorded phenotypes for each individual; α

is a vector indicating the intercept plus the considered fixed effects, i.e., batch effect

with 4 categories (all traits), farm of origin effect with 3 categories (all traits) and

laboratory of processing with 2 categories (GM and liver microarray expression data).

The α vector also includes the following covariates: IMF of LD (for LD fatty acid

composition traits), IMF of GM (for GM fatty acid composition traits), live weight (for

backfat thickness) and carcass weight (for ham fat thickness); W corresponds to the

incidence matrix relating phenotypes with their corresponding effects; x is the genotype

vector for the 12 miRNA polymorphisms; δ is the allele substitution effect for each

polymorphism; u is a vector of random individual genetic effects with an n-dimensional

multivariate normal distribution MVN_n (0, λ τ^{-1} K), where τ^{-1} corresponds to the

variance of the residual errors, λ is the ratio between the two variance components and

K is the known relatedness matrix derived from SNP information; and ε is the vector of

residual errors.

Association analyses were performed between each miRNA polymorphism

(**Table 1**) and the 38 lipid-related traits listed in **Additional file 3: Table S3**. Moreover,

we explored the associations between miRNA SNPs and expression levels of mRNAs

that fulfilled the following conditions: 1) the 3'UTR of the mRNA contains at least one

7mer-m8 site complementary to the seed of the miRNA harboring the SNP to be tested

(as predicted with the *locate* tool from SeqKit software [46]), 2) the existence of

experimentally validated mRNA-miRNA interactions between the given mRNA and the

tested miRNA have been confirmed in humans according to information provided in the

DIANA-Tarbase v8 database [63]. The first condition was established because we were

interested in focusing our analysis on miRNA-mRNA pairs for which a direct physical

interaction (through complementarity between the seed of the miRNA and a target site

in the 3'UTR of the mRNA) is feasible. The second condition was imposed because

there is extensive evidence that in silico prediction or miRNA binding sites can have

high false positive rates [64]. We considered as valid miRNA-mRNA interactions those

experimentally validated in humans by using wet lab methods such as cross-linking,

ligation and sequencing of hybrids (CLASH), photoactivatable ribonucleoside-enhanced

and   high-throughput   sequencing   of   RNA   isolated   by crosslinking   and

immunoprecipitation (PAR-CLIP and HITS-CLIP) and luciferase assays [63].

Importantly, the seeds of the porcine miRNAs harboring the 12 genotyped SNPs

were completely conserved with regard to the seeds of the human orthologous miRNA

genes (**Additional file 4: Fig. S1**). This feature further supports the extrapolation of

miRNA-mRNA interactions experimentally validated in humans [63] to pigs. For the

sake of completeness, we also carried out a complementary association analysis

between miRNA SNPs and the mRNA levels of the whole sets of genes expressed in the

GM and liver tissues (i.e., without applying the two conditions mentioned before) but

the results of such analyses should be interpreted with caution due to reasons stated

above.

In association analyses with lipid traits or gene expression phenotypes, the

alternative hypothesis $H_1 : \delta \neq 0$ was contrasted against the null hypothesis $H_0 : \delta = 0$

with a likelihood ratio test. The statistical significance of the associations between

miRNA SNPs and lipid and mRNA expression phenotypes was assessed with a false

discovery rate (FDR) approach [65].

Pathway enrichment analysis

The lists of probes/genes significantly associated at the nominal level (*P*-value < 0.05)

with miRNA SNPs genotypes (N = 12) after considering the criteria for miRNA target

inference (see previous Methods section) were used as inputs for pathway enrichment

analyses. The ClueGO v2.5.0 plug-in application [66] within the Cytoscape 3.8.2

software [67] was used for determining enriched Reactome and KEGG pathways. A

one-sided hypergeometric test of significance was applied for determining enriched

terms and multiple testing correction was implemented with a false discovery rate

approach [65].

Confirming associations between ssc-miR-326 rs319154814 genotypes and gene

expression data by RT-qPCR

The hepatic mRNA levels of three of the mRNA transcripts amongst the most

significantly associated with ssc-miR-326 rs319154814 (G/A) genotype were analyzed

by reverse transcription-quantitative polymerase chain reaction (RT-qPCR). The β-actin

(*ACTB*) and TATA-Box binding protein (*TBP*) genes were used as endogenous

controls, as previously reported [68]. In brief, 100 mg of liver tissue from 10 selected

samples (5 from each GG and AA genotypes) were submerged in 1 mL TRIzol reagent

(Invitrogen Corp., Carlsbad, CA, United States) and subsequently homogenized using

the Lysing Matrix D reagent (MP Biomedicals, Santa Ana, CA) in a Precellys 24 tissue

homogenizer (Bertin Instruments, Rockville, MD). Total RNA isolation was performed

following the protocol described by Rio et al. (2010) [69]. Reverse transcription was

achieved with the High-Capacity cDNA Reverse Transcription Kit (Applied

Biosystems, Foster City, CA) following the instructions of the manufacturer. We used 1

µg of total RNA as a template in a final volume of 20 µl and the synthesized cDNA was

diluted to 1:20 in Milli-Q water. One pair of primers spanning exon-exon junctions

were designed for each gene (**Additional file 5: Table S4**) with the Primer Express

software (Life Technologies Corporation). Real-time qPCR reactions contained 10 µl of

SYBR Select Master Mix, 300 nM of each primer, and 5 µl of 1:20 cDNA dilution, in a

final volume of 20 µl. Reactions were run in an ABI PRISM 7900HT instrument

(Applied Biosystems, Foster City, CA). The thermal cycle was set as follows: 2 min at

50 ºC, one denaturing step at 95 ºC during 10 min, followed by 40 cycles of 15 sec at 95

⬜C and 1 min at 60 ºC. Moreover, a melting curve analysis was performed (i.e., 95 °C

for 15 sec, 60 °C for 15 sec and a gradual increase in temperature, with a ramp rate of

1% up to 95 °C, followed by a final step of 95 °C for 15 sec), in order to assess the

specificity of the reactions. A standard curve with serial dilutions from a pool including

all the cDNA samples was used to assess whether amplification efficiencies of the three

genes under analysis were comprised in the 90 to 110% range. All reactions were run in

triplicate.

The $2^{-\Delta\Delta Ct}$ method [70] was used for the relative quantification of gene expression (Rq),

using the group of GG samples as a calibrator. Subsequently, Rq values were $\log_2$

transformed and the significance of expression differences between GG and AA

genotypes was assessed using Welch's t-test for unpaired groups of samples [71]

implemented in the *t.test* R function.

Measuring the expression of the ssc-miR-326 gene in pigs with different rs319154814

genotypes.

Liver samples from 9 AA and 9 GG pigs (rs319154814 genotype) were used to measure

the expression of the ssc-miR-326 gene. Total RNA was isolated with TRIzol as

previously described and reverse transcribed with the TaqMan Advanced MicroRNA

Reverse Transcription Kit (Applied Biosystems, Foster City, CA), following the

instructions of the manufacturer. Total RNA (10 ng) diluted in a volume of 2 μl, was

used as a template to generate and pre-amplify cDNA by carrying out four consecutive

reactions (poly-A tailing, adaptor ligation, reverse transcription and pre-amplification).

The resulting pre-amplified cDNA was diluted to 1:50 in Milli-Q water. To measure

ssc-miR-326 expression levels, both let-7a and miR-26a-5p were chosen as endogenous

controls following Timoneda et al. (2012) [72]. To conduct the experiments for the

three miRNAs under investigation (one target and two controls), available inventoried

TaqMan advanced miRNA assays (Applied Biosystems, Foster City, CA) for each

miRNA were purchased. The reactions were performed in a final volume of 15 μl that

contained: 3.75 μl of 1:50 cDNA, 3 μl of nuclease-free water, 7.5 μl of TaqMan Fast

Advanced Master Mix (2×), and 0.75 μl TaqMan Advanced miRNA Assay (20×). Each

reaction was carried out in triplicate in a 384-well plate. Reactions were run in a

QuantStudio 12K Flex Real-Time PCR System (Applied Biosystems, Foster City, CA).

The thermal cycle program was the same used for qPCR experiments on mRNA

transcripts. No standard dissociation curve was performed. The results were analyzed

applying the $2^{-\Delta\Delta Ct}$ method [70] for the relative quantification of miRNA expression

(Rq), using the group of GG samples as a calibrator. Subsequently, Rq values were $\log_2$

transformed and the significance of expression differences between GG and AA

genotypes was assessed using a Welch's t-test for unpaired groups [71].

Results

Differential segregation of SNPs mapping to microRNA genes in pigs and wild

boars from Europe and Asia

A total of 58,537,491 million SNPs were identified with the GATK haplotype

caller tool [41] in a data set comprising 120 WGS from 40 EDM, 40 ADM, 20 EWB

and 20 AWB individuals (**Additional file 1: Table S1**). The distribution of these SNPs

within different annotated regions (i.e. protein coding genes, exons, introns, 3'UTRs,

miRNAs, lncRNAs, snRNAs, snoRNAs and pseudogenes) can be found in **Additional**

file 6: Table S5. The majority of these SNPs were biallelic (96.82%), 777,008 (1.33%)

were tri-allelic, 1,864,388 (3.18%) had a deletion allele and 57,878 (0.098%) showed

four or more alleles (multi-allelic). From the set of multi-allelic SNPs, 1,765 SNPs

mapped to exons, and 531 of them were clustered in a set of 74 genes enriched in

olfactory receptors (**Additional file 7: Table S6**). This latter result is consistent with the

potential presence of copy number variation (CNV) in such locations [73]. Alternative

allele frequencies were consistently high (> 0.5) for 9.25% of variants, whereas low

(between 0.05 and 0.01) and very low (< 0.01) alternative allele frequencies were

detected in 27.85% and 19.62% of SNPs, respectively.

After filtering, 19,720,314 autosomal whole-genome SNPs were selected for

assessing population structure based on PCA clustering techniques. The PCA showed a

strong genetic differentiation amongst Asian and European populations (**Fig. 2A**). In

contrast, domestic pigs and wild boars, and particularly those with European origin, did

not show such stark divergence. Asian pigs and wild boars displayed some level of

genetic differentiation and were more diverse than their European counterparts.

With regard to miRNA variability, the 370 porcine miRNA genes annotated in

the manually curated miRCarta v1.1 database [44] were selected, and SNPs within these

genes were retrieved. A total of 285 SNPs residing in 139 miRNAs (37.56% of the total

count) were identified (**Additional file 2: Table S2**), implying that most of miRNAs are

monomorphic. The majority of these 139 miRNA loci (76.98%) contained 1-2 SNPs

located inside their predicted genomic boundaries, while 18.70% contained between 3

up to 5 variants, and 4.32% of them displayed more than 7 SNPs (**Additional file 8:**

Fig. S2). Only 43 miRNA SNPs (15.09%) were shared amongst EDM, ADM, EWB and

AWB populations (**Additional file 2: Table S2**), showing alternative alleles in at least

one of the analyzed individuals in each group. The number of SNPs segregating in each

of the four defined groups were 129 (EDM), 201 (ADM), 76 (EWB) and 172 (AWB),

respectively (**Additional file 2: Table S2**). With regard to precursor and mature

regions, 41 and 2 SNPs were shared amongst the four populations under consideration,

respectively (**Additional file 9: Fig. S3A** and **S3B**). None of the SNPs in the seed

regions were shared by the four porcine populations (**Additional file 9: Fig. S3C**). Only

three miRNA SNPs were found in the European data set but not in the Asian one. In

strong contrast, 55 miRNA SNPs were detected in the Asian data set but not in the

European one. Principal component analyses based on identified autosomal miRNA

SNPs (N = 260, **Additional file 2: Table S2**) showed the existence of a poor

differentiation between pigs and wild boars (**Fig. 2B**), while the genetic divergence

between European and Asian individuals was still apparent.

When we analyzed the population structure based on whole-genome autosomal

3'UTR SNPs (N = 709,343 SNPs, **Fig. 2C**), 3'UTR 7mer-m8 site SNPs (N = 107,196

SNPs, **Fig. 2D**) and 3'UTR 8mer site SNPs (N = 33,511 SNPs, **Fig. 2E**), the genetic

differentiation between Asian vs European populations was evident, in close

concordance with results shown in **Fig. 2A** and **2B**. However, we also detected a more

pronounced differentiation between domestic pigs and wild boars, and this observation

was particularly applicable to Asian pigs and wild boars (**Fig. 2C-D**).

The analysis of European and Asian populations shows reduced variability in

porcine microRNAs

About 47.76%, 57.36%, 44.77% and 36.84% of miRNA SNPs showed

alternative allele frequencies ≤ 0.1 in the ADM, EDM, AWB and EWB populations,

respectively (**Fig. 3, Additional file 2: Table S2**). Variations located at mature miRNA

and seed regions were enriched in rare or very rare variants when compared to the

variability of miRNA precursor regions (**Additional file 2: Table S2**), with average

alternative allele frequencies of ~0.1 for mature and seed miRNA polymorphisms. In

contrast, the average alternative allele frequency observed for SNPs in precursor areas

was ~0.15.

Moreover, the observed SNP density adjusted to 100 bp for miRNA precursor,

mature and seed regions consistently followed the order *precursor > mature > seed*

when we considered the full set of 120 WGS. Indeed, ~1 SNP per 100 bp was detected

in precursor regions, whereas ~0.86 and ~0.42 SNPs per 100 bp were observed in the

mature and seed regions, respectively (**Fig. 4A**). After computing a non-parametric

comparison of SNP density between the mature and seed regions with a Mann Whitney

U test [50], a significant difference (*P*-value = 0.016) was detected. These results

implied a 1.16-fold reduction in SNP density between precursor and mature regions,

while for seed regions, which are critical determinants of miRNA-mRNA interaction,

the observed SNP density was ~2.4 fold lower than in precursor regions. This

differential distribution of the SNP density across miRNA regions (*precursor > mature*

> seed) was also observed in each of the analyses performed in the ADM, EDM, AWB

and EWB groups (**Fig. 4A**). With regard to variants located within mature miRNAs (N

= 64), both inside (N = 12) and outside seed regions (N = 52), their distribution along

the sequence of the mature miRNA (~22 nts) showed a characteristic pattern (**Fig. 4B**):

amongst all the detected SNPs, the 1st position of the mature miRNA 5' end, which

binds to the MID domain of the Argonaute protein in the miRISC complex, exhibited a

SNP density of ~0.49 SNPs per 100 bp. Such scarcity in polymorphic sites was also

observed when considering the next 2nd to 8th positions in the mature miRNA sequence

(seed region), with an average of ~0.42 SNPs per 100 bp across the whole seed and up

to ~0.73 SNPs per 100 bp in the 6th position of the mature miRNA. In contrast, the

interval comprising positions 9th to 12th (with no functional implications in terms of

mRNA targeting) showed an increased average SNP density of ~0.98 SNPs per 100 bp.

Positions 13th to 18th of the mature miRNA, which roughly correspond to a functional

region providing additional anchor pairing to the seed region, showed a decreased SNP

density, particularly at positions 16th to 17th (**Fig. 4B**). Additionally, an increased SNP

density was found at positions 19th to 22nd. Furthermore, the SNP density in miRNA

genes (N = 370, **Fig. 4C**) was ~2.6 fold lower (*P*-value < 0.01, see Methods) than in

their flanking regions (± 1 kb). A list of the miRNA SNPs (N = 64) located at mature

miRNAs and their genomic coordinates within the mature miRNA sequence is available

at **Additional file 10: Table S7.**

Statistics of the whole-genome sequencing of five Duroc boars

As previously explained, we sequenced the genomes of five Duroc pigs which

founded a population of 350 offspring with the goal of identifying SNPs in miRNA

genes and investigating their association with mRNA expression and lipid phenotypes.

Mean coverage values of the five pig genomes ranged from 37.67× to 46.6×, with more

than 98.6% of the genome covered by at least 10 reads in all five samples, and 96.71%

of the genome covered by at least 15 reads. More details about coverage and genome

mapping parameters are shown in **Additional files 11** and **12: Table S8** and **Fig. S4**.

After performing variant calling on mapped reads, a total of 13,839,422 SNPs passed

the established quality filters, whereas 3,721,589 insertions and deletions (INDELs)

were detected. Moreover, a total of 1,643,861 INDELs (44.17%) and 6,034,548 SNPs

(43.60%) were located in annotated protein coding loci. A total of 55 SNPs and 5

INDELs were located in the 370 miRNA loci annotated in the Sscrofa.11.1 genome

assembly according to the miRCarta database [44]. From these, 53 variants (96.36%)

were shared with the set of 285 miRNA SNPs found in the 120 WGS reported before

(**Additional file 13: Table S9**). After performing quality check of the variants mapping

to miRNA genes, 49 SNPs and 5 INDELs passed all established filtering criteria.

Amongst the 49 identified miRNA polymorphisms, we selected 12 SNPs on the

basis of their location at relevant annotated miRNA loci (**Table 1, Additional file 13:**

Table S9) for genotyping and performing association analyses. Microarray expression

data comprised a total of 23,998 probes, of which 12,131 (50.55%) and 12,787

(53.28%) corresponded to genes expressed in GM and liver tissues, respectively.

To provide a comprehensive view about the association between miRNA

genotypes and liver and muscle mRNA levels, we have carried out an association

analysis between miRNA genotypes and the whole set of expressed probes/genes in

these two tissues (**Table 2**). It can be seen that there are hundreds of associations at the

nominal level, while the number of significant associations is strongly reduced after

multiple testing correction. This analysis does not take into account neither the

existence of complementarity between the seed of the miRNA and the porcine 3'UTR

of target mRNAs, nor functional validation data obtained in humans (please see the

Methods section). A second analysis taking into account these two criteria revealed that

from the set of 12 genotyped SNPs listed in **Table 1**, only 2 SNPs showed significant

associations (q-value < 0.1) with the levels of mRNAs predicted to be miRNA targets

(**Table 3, Additional file 14: Table S10**). Further details of target sites at the 3'UTR of

mRNA transcript expression profiles significantly associated with miRNA SNP

genotypes (**Table 3**) are available at **Additional file 15: Table S11**). All genes

displaying significant associations (q-value < 0.1) in this second analysis ranked

amongst the top 1% of significance (**Additional file 16: Table S12**) in the condition-

free analysis (Analysis 1), except for transcripts associated with ssc-miR-326 in the

liver, which ranked amongst the top 5% (**Table 1, Additional file 16: Table S12**). We

observed that for several of the significant associations between miRNA and mRNA

pairs reported in **Table 3** and **Additional file 14: Table S10**, the interaction was

classified as 7-mer-m8 by our targeting approach with the *locate* tool of SeqKit toolkit

[46] when, in reality, it was of type 8mer with one mismatch (data not shown).

Two SNPs in the apical loop of ssc-miR-326 and in the precursor region of ssc-miR-23a are associated with the mRNA expression of several of their putative targets.

When we analyzed the association between the rs319154814 (G/A) polymorphism located in the apical loop of ssc-miR-326 and gene expression data (**Table 3**), significant results were obtained after multiple testing correction (*q*-value < 0.1). More specifically, we detected eight significant associations between this SNP and the hepatic mRNA expression of experimentally confirmed targets of this miRNA (**Table 3**). For instance, the hepatic mRNA levels of the cellular FLICE-like inhibitory protein (*CFLAR*), the protein phosphatase 1 catalytic subunit γ (*PPP1CC*), the syndecan 1 (*SDC1*), the splicing factor 3A subunit 3 (*SF3A3*) or the Follistatin-like 1 (*FSTL1*) were amongst the most significantly associated with ssc-miR-326 genotypes (**Table 3**). The expression levels of seven of these mRNAs associated with ssc-miR-326 rs319154814 (G/A) genotypes (**Table 3**) were reduced in pigs homozygous for the mutated allele (AA, N = 32), as depicted in **Fig. 5**.

The mRNA levels of the *PPP1CC*, *CFLAR* and *SF3A3* mRNA transcripts were measured by RT-qPCR in 10 liver samples (5 belonging to each AA or GG rs319154814 genotypes). After comparing the mean expression for each of the three analyzed mRNA transcripts in both AA and GG genotype groups, only the *PPP1CC* transcripts showed a significantly reduced expression in AA pigs compared with their GG counterparts (*P*-value = 0.027, **Fig. 6A**). The other two transcripts (*CFLAR* and *SF3A3*) also showed a reduced expression in pigs with AA genotype (**Fig. 6A**), but results were not significant. We further investigated whether the ssc-miR-326 rs319154814 (G/A) genotypes were associated with the levels of this very same miRNA in the liver by using a specific Taqman Advanced miRNA probe assay. In this qPCR

analysis, pigs with AA genotypes for the rs319154814 polymorphism showed an increased ~1.9-fold expression of ssc-miR-326 transcripts compared with their GG counterparts, although such difference did not reach statistical significance (P-value = 0.178, **Fig. 6B**).

The rs333787816 (T/C) polymorphism, located in the precursor region, immediately downstream the mature ssc-miR-23a sequence, was also significantly associated with 36 experimentally confirmed targeted genes in the GM muscle tissue (q-value < 0.1, **Table 3, Additional file 14: Table S10**). From these, it is worth mentioning the Argonaute RISC component 1 (*AGO1*) and the Argonaute RISC catalytic component 2 (*AGO2*) mRNAs. Both *AGO1* and *AGO2* genes showed lower mRNA expression levels in homozygous CC pigs with respect to their TT and TC counterparts (**Table 3**).

Pathway enrichment analyses of significantly associated mRNA targets provide insights of putative metabolic functions of miRNAs.

After performing pathway enrichment analyses on the sets of putative targeted mRNA transcripts significantly associated with miRNA SNP genotypes (P-value < 0.05), several biological categories were identified and reported at **Additional file 17: Table S13**. To mention a few of the most relevant findings, the ssc-miR-23a genotype was associated with several regulatory pathways (post-transcriptional silencing by small RNAs, competing endogenous RNAs regulate PTEN translation, regulation of PTEN mRNA translation, regulation of RUNX1 Expression and Activity etc.) involving genes, such as *AGO1*, *AGO2* or *TNRC6C* genes, which play relevant roles in miRNA-mediated gene regulation [74]. Amongst the sets of mRNAs nominally associated with ssc-miR-

30d genotype in GM, we identified the regulation of ornithine decarboxylase activity or

cellular response to hypoxia, as well as mature mRNA transport after splicing

(**Additional file 17: Table S13**).

In the liver tissue, the ssc-miR-326 genotype was significantly associated with the

expression of several circadian clock regulatory transcripts e.g., histone deacetylase 3

(*HDAC3*), period 1 (*PER1*) and the already mentioned *PPP1CC* gene. Moreover,

negative regulation of DDX58/IFIH1 signaling of type I interferon response and p53

signaling were pathways enriched in the sets of mRNAs nominally associated with ssc-

miR-130a genotypes (**Additional file 17: Table S13**).

Porcine lipid phenotypes are associated with the genotypes of miRNA genes

We also evaluated the association between miRNA SNPs and several lipid-

related phenotypes recorded in the Lipgen population (**Additional file 3: Table S3**).

Only the rs319154814 (G/A) variant in the ssc-miR-326 gene was significantly

associated (*q*-value < 0.1) with lipid traits (**Table 4**, **Additional file 18: Table S14**).

More specifically, we found significant associations with myristic acid (C14:0) content

in both LD and GM muscles, as well as with the gadoleic acid (C20:1) content and the

ratio between PUFA and MUFA in the LD muscle (**Table 4, Additional file 18: Table**

S14). We also observed several associations between the rs319154814 (G/A) SNP and

fatty acid composition traits, but they were significant only at the nominal level (*P*-

value < 0.01, **Table 4**). Other apical loop SNPs like rs81349391 (A/G) at ssc-miR-339

and rs344472188 (T/C) at ssc-miR-130a were significantly associated at the nominal

level (*P*-value < 0.01) with palmitic acid (C16:0) content and SFA and UFA proportion

in the GM muscle, as well as with backfat thickness. As shown in **Table 4**, a SNP

located in the precursor 3' stem of ssc-miR-30d showed a nominally significant

association with the content of arachidic acid (C20:0). Other significant associations at

the nominal level were, for instance, those between the rs334590580 (T/C) SNP located

at the precursor stem region of ssc-miR-335 and palmitic and arachidic acids content in

GM tissue (**Table 4**).

Discussion

Divergent patterns of variation for microRNA and 3'UTR polymorphisms in Asian

and European pigs and wild boars

The PCA revealed the existence of a detectable genetic differentiation between

Asian and European populations, with the latter showing reduced levels of diversity

when compared to the former. Groenen et al. (2012) [75] investigated the variability of

pig genomes and found that Asian pigs and wild boars are more diverse than their

European counterparts and that both gene pools split during the mid-Pleistocene 1.6–0.8

Myr ago. Calabrian glacial intervals probably favored a restricted gene flow between

these two gene pools [75]. The high variability of Asian populations could be explained

by the fact that *Sus scrofa* emerged as a species in Southeast Asia (5.3-3.5 Mya) and

then dispersed westwards until reaching Europe around 0.8 Mya [76]. This initial

founder effect combined with the occurrence of strong bottlenecks reduced the genetic

diversity of European wild boars [75].

While genetic differentiation between wild boar and pig populations was clearly

discernible considering the whole-genome SNP data set (**Fig. 2A**), this was less evident

in the PCA based on miRNA SNPs (**Fig. 2B**), probably because the low number (285

SNPs) of markers employed in this analysis limits the resolution with which population

differentiation can be detected. We have also found that the degree of population

differentiation between Asian domestic pigs and Asian wild boars increases when PCAs

are built on the basis of SNPs located in the 3'UTR (709,343 SNPs), 3'UTR 7mer-m8

sites (107,196 SNPs) or 3'UTR 8mer (33,511 SNPs) sites (**Figs. 2C-E**). The potential

effects of 3'UTR SNPs are the modulation of mRNA expression, secondary structure,

stability, localization, translation, and binding to miRNAs and RNA-binding proteins

[77], so in general, they are not expected to have drastic consequences on gene function

[77]. Purifying selection is less intense in 3'UTRs than in protein-coding regions,

implying that 3'UTRs evolve faster and accumulate a larger fraction of recent

polymorphisms contributing to population differentiation [78].

Low SNP density in microRNA genes and lack of uniform SNP distribution across

sites

We have found that, in general, miRNA loci have a substantially lowered SNP

density in their seeds when compared with mature and precursor regions (**Fig. 4A**).

When we analyzed the SNP density in miRNA loci and their flanking regions (± 1 kb),

a significantly reduced amount of SNPs per 100 bp was found in miRNAs compared

with upstream and downstream flanking sequences (**Fig. 4C**). These results were in

agreement with data reported by Omariba et al. (2020) [79] and Saunders et al. (2007)

[5].

The low variability of miRNA genes, a feature that was particularly evident in

the seed region, is probably due to the intense effects of purifying selection. Indeed, the

importance of the miRNA seeds is revealed by the high conservation of their sequence

across species [16,23,80], as this sequence ultimately determines the success of

miRNA-mRNA interactions [2]. In our study, a total of 221, 52 and 12 SNPs were

found in precursor, mature and seed regions within miRNA loci, respectively

(**Additional files 2** and **9: Table S2, Fig. S3**). Gong et al. (2012) [81] described the

existence of 40% polymorphic miRNAs in the human genome but only 16% of them

displayed more than one SNP. In a more recent study, He et al. (2018) [82] reported

1,879 SNPs in 1226 (43.6%) human miRNA seed regions, and 97.5% of these

polymorphisms had frequencies below 5%. These results agree well with the overall

frequency distribution of miRNA SNPs in the European and Asian populations analyzed

in the current work (**Fig. 3**). He et al. (2018) also demonstrated that 1,587, 749, 340,

102, 31, and 4 miRNAs harbored zero, one, two, three, four, and five SNPs,

respectively, in their seed regions, reflecting that mutations in this critical functional

region are not well tolerated [82]. This distribution is similar to the one that we have

observed in domestic pigs and wild boars, with 81, 31, 11, 9 and 5 miRNAs harboring

one, two, three, four and five SNPs (**Additional file 8: Fig. S2**). Only 4 and 2 miRNAs

showed a total of seven and ten polymorphisms within their sequences (**Additional file

2: Table S2**).

We have also detected a non-uniform SNP density along the sequence of mature

miRNAs (**Fig. 4B**). Gong et al. (2012) [81] showed that SNPs tend to concentrate in the

middle region of the mature miRNA gene rather than in its 5' and 3'ends. Moreover, the

same authors described an increased SNP density at positions 9 and 15 of the mature

miRNA, a result that agrees well with ours (**Fig. 4B**) However, we have also identified

an elevated number of polymorphic sites at positions 11, 19 and 20, a finding that does

not match human data presented by Gong et al. (2012) [81].

Several of the sites showing a reduced variability in porcine miRNAs exert

critical functions (**Fig. 4B**, **Additional file 10: Table S7**). For instance, the 1st

nucleotide of mature miRNAs plays an important role in the loading process of the

mature miRNA on the Argonaute protein to form the miRISC complex [83].

Nucleotides 2nd to 8th in the mature miRNA correspond to the seed, where we found a

consistently reduced SNP density (**Fig. 4A** and **4B**) compared with other miRNA

regions. This result was expected because this region has a crucial role in modulating

the interaction between the mature miRNA and its 3'UTR binding sites. Polymorphisms

in the seed region have the potential to disrupt the proper miRNA-mRNA pairing and

thus alter biologically relevant regulatory pathways, which tend to be evolutionarily

conserved [80]. Hence, these variants located in the miRNA seeds (**Additional file 10:**

Table S7) might favor the emergence of novel miRNA-mRNA interactions as well as

the abolishment of conserved ones, thus modifying gene regulatory networks.

In contrast with the first eight 5' positions of the miRNA, we found a higher

SNP density in positions 9th to 12th, which do not contribute substantially to miRNA

target recognition [2] (**Fig. 4B**). Positions 13th to 16th facilitate 3'-compensatory pairing

between the mature miRNA and targeted 3'UTRs [84], although only at marginal levels

[85]. Nevertheless, in our porcine data set only positions 16th-17th showed a reduced

SNP density (**Fig. 4B**).

Polymorphisms in microRNA genes show associations with the mRNA expression

of several of their predicted targets

We have investigated the association of 12 SNPs mapping to miRNA genes and

hepatic and muscle mRNA expression in Duroc pigs. Several hundreds of significant

associations at the nominal level (P-value < 0.05) were detected in both GM muscle and

liver tissues when considering the whole set of microarray expression profiles of mRNA

transcripts (**Table 2**). One pitfall of this analysis is that it does not take into account

whether there is molecular evidence supporting the existence of direct miRNA-mRNA

interactions displaying significant associations. On the other hand, indirect interactions

between miRNAs and mRNAs might also exist i.e., a SNP regulating the expression of

a miRNA might repress the translation of a given mRNA through a direct interaction

(an event that would not necessarily imply any transcriptional decay, hence being

undetectable in our experimental system) and, in turn, this might indirectly affect the

expression of other multiple mRNAs regulated by the translationally repressed mRNA.

Despite this consideration, we decided to restrict our association analyses to miRNA-

mRNA pairs with molecular evidence of potential interactions because we believe that

with this stringent approach, we are able to minimize the chances of detecting spurious

false positive results. The genomic position of miRNA-binding sequences in the 3'UTR

of targeted mRNAs showing significant associations (q-value < 0.1, **Table 3**) with

miRNA genotypes is available in **Additional file 15: Table S11**).

After correction for multiple testing (q-value < 0.1, **Table 3**), only SNPs

mapping to the ssc-miR-23a and ssc-miR-326 genes displayed significant associations

with the expression levels of some of their putative mRNA targets. More in detail, the

rs333787816 (T/C) polymorphism in the ssc-miR-23a gene showed significant

associations with several putative target mRNAs, amongst which are *AGO1* and *AGO2*

transcripts, two essential components of the miRNA-mediated silencing machinery

[86,87]. The rs319154814 (G/A) polymorphism in the apical loop region of ssc-miR-

326 also showed a significant association, after multiple testing correction, with the

hepatic mRNA expression of several of its predicted and experimentally confirmed

targets (**Table 3**). In contrast, no association with GM muscle mRNA expression was

observed. Tissue-specific differences in the expression of the miRNA or its mRNA

targets might explain such outcome. Indeed, the analysis of the distribution of miRNA

expression across human tissues has shown that only a minority of miRNAs are

expressed ubiquitously [88]. The hepatic mRNA targets showing the most significant

association with rs319154814 (G/A) ssc-miR-326 genotype were *CFLAR*, *PPP1CC,*

SDC1, SF3A3 and *FSTL1*. The protein encoded by the *PPP1CC* gene belongs to the

protein phosphatase PP1 subfamily, which is a ubiquitous serine/threonine phosphatase

involved in regulating multiple cellular processes through dephosphorylation signaling.

Amongst them, it is worth mentioning insulin signaling [89], post-translational

localization of circadian clock components [90] and lipids [91,92] or glycogen

metabolism regulation [93]. Moreover, the cellular FLICE-like inhibitory protein

(*CFLAR*) gene encodes the cFLIP protein and is involved in the inhibition of Fas-

mediated apoptosis [94], while the Follistatin-like 1 (*FSTL1*) gene plays a role in the

immune inflammatory signaling and fibrosis in the liver [95].

Pigs homozygous for the derived A-allele of the rs319154814 SNP showed a

consistent downregulation of the hepatic mRNA expression of the *CFLAR*, *PPP1CC,*

SDC1, SF3A3, FSTL1, NAA50 and *ELAVL1* genes (**Fig. 5**), a result that was confirmed

by qPCR for *PPP1CC* mRNA transcripts (*P*-value = 0.027, **Fig. 6A**). Although in the

qPCR experiment the *CFLAR* and *SF3A3* genes also showed decreased mRNA levels in

pigs with homozygous AA genotypes for the rs319154814 SNP, such associations did

not reach statistical significance (**Fig. 6A**). The significant downregulation of at least

one top mRNA putative target (i.e., *PPP1CC*) might suggest that the rs319154814

(G/A) variant could increase the repressive activity of ssc-miR-326. Indeed, when we

assessed, with a dedicated qPCR assay, the expression level of ssc-miR-326 transcripts

in the liver tissue, we found that ssc-miR-326 $\log_2$ relative expression is increased by

~1.9 fold in pigs homozygous for the A-allele, although such result was not statistically

significant (*P*-value = 0.178, **Fig. 6B**).

817   As previously discussed, the rs319154814 (G/A) polymorphism is located in the

apical loop region of ssc-miR-326. Although the apical region of the miRNA does not

have a function as critical as the seed, polymorphisms located in this specific region can

have relevant effects on the structural conformation of the pre-miRNA hairpin. Indeed,

Fernandez et al. (2017) [19] described a mutation in the apical loop of hsa-miR-30c

(G/A) that induces a steric disruption of the pri-miRNA folding structure of the hairpin,

hence creating a bulge around the flanking downstream CNNC motif that facilitates the

SRSF3 factor accessibility to the RNA sequence [96]. In other words, SNPs in the

apical region can modify the efficiency with which the Drosha processing machinery, is

recruited. The rs319154814 (G/A) polymorphism detected in our study might have

structural consequences similar to those described for the hsa-miR-30c apical loop

variant [19].

829   One clear limitation of our experimental design is that we did not test the

existence of miRNA-mRNA interactions by using functional assays. Such experiments

would be necessary to demonstrate our hypothesis that the rs319154814 (G/A)

polymorphism in the apical loop of ssc-miR-326 has a role in promoting the processing

and maturation efficiency of the hairpin precursor, thus modifying the expression

profiles of ssc-miR-326 and its putative mRNA targets.

A polymorphism in the apical loop of microRNA 326 is associated with fatty acid

composition traits

In our study, the only miRNA polymorphism showing significant associations

(q-value < 0.1) with lipid-related phenotypes was the rs319154814 (G/A) SNP in the

apical loop of ssc-miR-326. This SNP was associated with myristic fatty acid content in

the GM and LD muscles of Duroc pigs, as well as with PUFA/MUFA ratio or gadoleic

fatty acid content in LD. To the best of our knowledge, no direct effect of ssc-miR-326

on the metabolism of myristic fatty acid has been described so far, but there are reports

suggesting that several of the targets of this miRNA might be involved in carbohydrate

and lipid metabolism pathways [97,98]. Moreover, the *PPP1CC* transcript, one of the

predicted targets of miR-326, encodes a subunit of protein phosphatase-1, which

activates acetyl-CoA carboxylase α (ACACA) and 6-phosphofructo2-kinase/fructose-

2,6-bisphosphatase (PFKFB), the main regulators of fatty acid synthesis and glycolysis,

respectively [92]. The protein phosphatase-1 also activates lipogenic transcription

factors such as sterol regulatory element-binding protein 1 (SREBF1), carbohydrate-

responsive element-binding protein (MLXIPL) and, moreover, it dephosphorylates the

DNA-dependent protein kinase encoded by the *PRKDC* gene, which is another main

determinant of hepatic lipogenesis [92]. In summary, a potential effect of the

rs319154814 (G/A) SNP in the synthesis or degradation of myristic acid can be

envisaged, but this hypothesis still needs to be confirmed at the functional level.

Interestingly, Cardoso et al. (2021) [99] explored the relationship between miRNA

SNPs located in the seed and mature regions of bovine miRNA loci and fatty acid

composition traits. By combining lipid-related phenotype data with mRNA and protein

abundances in a multi-omics approach, they described three miRNA SNPs significantly

associated with several unsaturated and polyunsaturated fatty acid traits, as well as with

the expression levels of some of their predicted targets [99].

Conclusions

MicroRNA genes show divergent patterns of variation between Asian and European pigs and wild boars and, in general, they display low levels of polymorphism. As expected, this reduced miRNA variability was particularly prevalent in the seed region, a finding that is likely explained by the strong effects of purifying selection aiming to preserve the conservation of this critical site. We have detected one SNP in the apical loop of ssc-miR-326 and another one in the precursor region of ssc-miR-23a that are associated with the mRNA expression of several of their putative targets. If confirmed with functional assays, these results would reinforce the need of exploring the role of miRNA variation within and outside the seed in the fine-tuning of mRNA expression in pigs.

Declarations

Ethics approval

Animal care and management procedures were carried out in accordance with national guidelines for Good Experimental Practices and they were approved by the Ethical Committee of the Institut de Recerca i Tecnologia Agroalimentàries (IRTA).

Consent of publication

Not applicable.

884   **Availability of data materials**

885   Microarray expression data used in the current study were deposited in the Gene

886   Expression Omnibus (GEO) public repository and are accessible through GEO Series

887   Accession Number GSE115484. Phenotypic and genotypic data sets generated and

888   analyzed during the current study have been deposited in the Figshare public repository

889   available   at   https://figshare.com/projects/SNPs_miRNA/78690.   Whole-genome

890   sequencing dataset from the 5 Duroc boars is available at the Sequence Read Archive

891   (SRA) database (BioProject: PRJNA626370). Variant Calling Format (VCF) files from

892   the 120 European and Asian pigs belonging to domestic breeds and wild boars, as well

893   as   from   the   5   sequenced   Duroc   boars   are   available   at   the   following   link:

894   https://figshare.com/projects/VCF_PIGs/93140.

895

896   **Competing interests**

897   The authors declare that they have no competing interests.

898

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911   Authors' contributions

MA conceived this study. MA and RQ designed the experimental protocols. RQ

coordinated phenotyping recording and contributed to generate microarray expression

data. EMS did DNA extractions and selected SNPs to be genotyped. EMS performed all

bioinformatic and statistical analyses of the data. MGL and AC conducted qPCR

analyses. DG contributed to population structure analyses. RT performed whole-

genome sequencing of the five Duroc boars. EMS and MA drafted the manuscript. All

authors read and approved the content of the final manuscript.

920   References

1. Gebert LFR, MacRae IJ. Regulation of microRNA function in animals. Nat Rev Mol

Cell Biol. 2019;20:21–37.

2. Bartel DP. Metazoan MicroRNAs. Cell. 2018;173:20–51.

3. Chen CYA, Shyu A Bin. Mechanisms of deadenylation-dependent decay. RNA.

2011;2:167–83.

4. Eichhorn SW, Guo H, McGeary SE, Rodriguez-Mias RA, Shin C, Baek D, et al.

MRNA destabilization is the dominant effect of mammalian microRNAs by the time

substantial repression ensues. Mol Cell. 2014;56:104–15.

5. Saunders MA, Liang H, Li WH. Human polymorphism at microRNAs and

microRNA target sites. Proc Natl Acad Sci U S A. 2007;104:3300–5.

6. Mencía A, Modamio-Høybjør S, Redshaw N, Morín M, Mayo-Merino F, Olavarrieta

L, et al. Mutations in the seed region of human miR-96 are responsible for

nonsyndromic progressive hearing loss. Nat Genet. 2009;41:609–13.

7. Auyeung VC, Ulitsky I, McGeary SE, Bartel DP. Beyond secondary structure:

primary-sequence determinants license pri-miRNA hairpins for processing. Cell.

2013;152:844–58.

8. Roden C, Gaillard J, Kanoria S, Rennie W, Barish S, Cheng J, et al. Novel

determinants of mammalian primary microRNA processing revealed by systematic

evaluation of hairpin-containing transcripts and human genetic variation. Genome Res.

2017;27:374–84.

9. Clop A, Marcq F, Takeda H, Pirottin D, Tordoir X, Bibé B, et al. A mutation creating

a potential illegitimate microRNA target site in the myostatin gene affects muscularity

in sheep. Nat Genet. 2006;38:813–8.

10. Moszyńska A, Gebert M, Collawn JF, Bartoszewski R. SNPs in microRNA target

sites and their potential role in human disease. Open Biol. 2017;7:170019.

11. Liang H, Li WH. Lowly expressed human MicroRNA genes evolve rapidly. Mol

Biol Evol. 2009;26:1195–8.

12. Xiong P, Schneider RF, Hulsey CD, Meyer A, Franchini P. Conservation and

novelty in the microRNA genomic landscape of hyperdiverse cichlid fishes. Sci Rep.

2019;9:1–12.

13. Peterson KJ, Dietrich MR, McPeek MA. MicroRNAs and metazoan

macroevolution: Insights into canalization, complexity, and the Cambrian explosion.

BioEssays. 2009;31:736–47.

14. Lee CT, Risom T, Strauss WM. Evolutionary conservation of microRNA regulatory

circuits: An examination of microRNA gene complexity and conserved microRNA-

target interactions through metazoan phylogeny. DNA Cell Biol. 2007;26:209–18.

15. Junjie LUO, Wang Y, Yuan J, Zhao Z, Jian LU. MicroRNA duplication accelerates

the recruitment of new targets during vertebrate evolution. RNA. 2018;24:787–802.

16. Simkin A, Geissler R, McIntyre ABR, Grimson A. Evolutionary dynamics of

microRNA target sites across vertebrate evolution. PLOS Genet. 2020;16:e1008285.

17. Li J, Zhang Z. MiRNA regulatory variation in human evolution. Trends Genet.

2013;29:116–24.

18. Sun G, Yan J, Noltner K, Feng J, Li H, Sarkis DA, et al. SNPs in human miRNA

genes affect biogenesis and function. RNA. 2009;15:1640–51.

19. Fernandez N, Cordiner RA, Young RS, Hug N, MacIas S, Cáceres JF. Genetic

variation and RNA structure regulate microRNA biogenesis. Nat Commun. 2017;8:1–

12.

20. Jin Y, Lee CGL. Single nucleotide polymorphisms associated with microRNA

regulation. Biomolecules. 2013;3:287–302.

21. Larson G, Cucchi T, Dobney K. Genetic aspects of pig domestication. In:

Rothschild MF, Ruvinsky A, editors. Genet of the pig. 2011. p.14–37.

22. Ramos-Onsins SE, Burgos-Paz W, Manunza A, Amills M. Mining the pig genome

to investigate the domestication process. Heredity. 2014;471–84.

23. Penso-Dolfin L, Moxon S, Haerty W, Di Palma F. The evolutionary dynamics of

microRNAs in domestic mammals. Sci Rep. 2018;8:1–13.

24. Wei W, Wang G, Cheng Y, Yang R, Song J, Huang S, et al. A miR-511-binding site

SNP in the 3ʹUTR of IGF-1 gene is associated with proliferation and apoptosis of PK-

15 cells. Vitr Cell Dev Biol – Anim. 2019;55:323–30.

25. Liu R, Deng D, Liu X, Xiao Y, Huang J, Wang F, et al. A miR-18a binding-site

polymorphism in *CDC42* 3ʹUTR affects *CDC42* mRNA expression in placentas and is

associated with litter size in pigs. Mamm Genome. 2019;30:34–41.

26. Ma C, Song H, Yu L, Guan K, Hu P, Li Y, et al. MiR-762 promotes porcine

immature Sertoli cell growth via the ring finger protein 4 (RNF4) gene. Sci Rep.

2016;6:32783.

27. Stachowiak M, Szydlowski M, Flisikowski K, Flisikowska T, Bartz M, Schnieke A,

et al. Polymorphism in 3ʹ untranslated region of the pig PPARA gene influences its

transcript level and is associated with adipose tissue accumulation. J Anim Sci.

2014;92:2363–71.

28. Bartz M, Szydlowski M, Kociucka B, Salamon S, Jeleń HH, Switonski M.

Transcript abundance of the pig stearoyl-CoA desaturase gene has no effect on fatty

acid composition in muscle and fat tissues, but its polymorphism within the putative

microRNA target site is associated with daily body weight gain and feed conversion

ratio. J Anim Sci. 2013;91:10–9.

29. Shao GC, Luo LF, Jiang SW, Deng CY, Xiong YZ, Li FE. A C/T mutation in

microRNA target sites in BMP5 gene is potentially associated with fatness in pigs. Meat

Sci. 2011;87:299–303.

30. Bartz M, Koscianska E, Szczerbal I, Nowacka-Woszuk J, Kociucka B, Salamon S,

et al. Polymorphism of the porcine miR-30d is associated with adipose tissue

accumulation, its fatty acid profile and the ME1 gene expression. Livest Sci.

2015;182:54–7.

31. Lee JS, Kim JM, Lim KS, Hong JS, Hong KC, Lee YS. Effects of polymorphisms in

the porcine microRNA MIR206 / MIR133B cluster on muscle fiber and meat quality

traits. Anim Genet. 2013;44:101–6.

32. Kim JM, Lim KS, Hong JS, Kang JH, Lee YS, Hong KC. A polymorphism in the

porcine miR-208b is Asso. with microRNA biogenesis and expressions of SOX-6 and

MYH7 with effects on muscle fibre characteristics and meat quality. Anim Genet.

2015;46:73–7.

33. Chai J, Chen L, Luo Z, Zhang T, Chen L, Lou P, et al. Spontaneous single

nucleotide polymorphism in porcine microRNA-378 seed region leads to functional

alteration. Biosci Biotechnol Biochem. 2018;82:1081–9.

34. Choi JW, Chung WH, Lee KT, Cho ES, Lee SW, Choi BH, et al. Whole-genome

resequencing analyses of five pig breeds, including Korean wild and native, and three

European origin breeds. DNA Res. 2015;22:259–67.

35. Kim H, Song KD, Kim HJ, Park W, Kim J, Lee T, et al. Exploring the Genetic

Signature of Body Size in Yucatan Miniature Pig. Johnson N, editor. PLoS One.

2015;10:e0121732.

36. Frantz LAF, Schraiber JG, Madsen O, Megens HJ, Cagan A, Bosse M, et al.

Evidence of long-term gene flow and selection during domestication from analyses of

Eurasian wild and domestic pig genomes. Nat Genet. 2015;47:1141–8.

37. Bosse M, Megens HJ, Madsen O, Crooijmans RPMA, Ryder OA, Austerlitz F, et al. Using genome-wide measures of coancestry to maintain diversity and fitness in endangered and domestic pig populations. Genome Res. 2015;25:970–81.

38. Bolger AM, Lohse M, Usadel B. Trimmomatic: A flexible trimmer for Illumina sequence data. Bioinformatics. 2014;30:2114–20.

39. Warr A, Affara N, Aken B, Beiki H, Bickhart DM, Billis K, et al. An improved pig reference genome sequence to enable pig genetics and genomics research. GigaScience. 2020;9:giaa051.

40. Li H. Aligning sequence reads, clone sequences and assembly contigs with BWA-MEM. 2013. http://arxiv.org/abs/1303.3997.

41. McKenna A, Hanna M, Banks E, Sivachenko A, Cibulskis K, Kernytsky A, et al. The genome analysis toolkit: A MapReduce framework for analyzing next-generation DNA sequencing data. Genome Res. 2010;20:1297–303.

42. O'Rawe J, Jiang T, Sun G, Wu Y, Wang W, Hu J, et al. Low concordance of multiple variant-calling pipelines: Practical implications for exome and genome sequencing. Genome Med. 2013;5:28.

43. Hasan MS, Wu X, Zhang L. Performance evaluation of indel calling tools using real short-read data. Hum Genomics. 2015;9:20.

44. Backes C, Fehlmann T, Kern F, Kehl T, Lenhof H-P, Meese E, et al. miRCarta: a central repository for collecting miRNA candidates. Nucleic Acids Res. 2018;46:160–7.

45. Kozomara A, Birgaoanu M, Griffiths-Jones S. miRBase: from microRNA sequences

to function. Nucleic Acids Res. 2019;47:155–62.

46. Shen W, Le S, Li Y, Hu F. SeqKit: A cross-platform and ultrafast toolkit for

FASTA/Q file manipulation. PLoS One. 2016;11:e0163962.

47. Grimson A, Farh KK-H, Johnston WK, Garrett-Engele P, Lim LP, Bartel DP.

MicroRNA targeting specificity in mammals: determinants beyond seed pairing. Mol

Cell. 2007;27:91–105.

48. Friedman RC, Farh KKH, Burge CB, Bartel DP. Most mammalian mRNAs are

conserved targets of microRNAs. Genome Res. 2009;19:92–105.

49. Price AL, Patterson NJ, Plenge RM, Weinblatt ME, Shadick NA, Reich D. Principal

components analysis corrects for stratification in genome-wide association studies. Nat

Genet. 2006;38:904–9.

50. Mann HB, Whitney DR. On a test of whether one of two random variables is

stochastically larger than the other. Ann Math Stat. 1947;18:50–60.

51. Vidal O, Noguera JL, Amills M, Varona L, Gil M, Jiménez N, et al. Identification of

carcass and meat quality quantitative trait loci in a Landrace pig population selected for

growth and leanness. J Anim Sci. 2005;83:293–300.

52. Gallardo D, Pena RN, Amills M, Varona L, Ramírez O, Reixach J, et al. Mapping of

quantitative trait loci for cholesterol, LDL, HDL, and triglyceride serum concentrations

in pigs. Physiol Genomics. 2008;35:199–209.

53. Gallardo D, Quintanilla R, Varona L, Díaz I, Ramírez O, Pena RN, et al.

Polymorphism of the pig *acetyl-coenzyme A carboxylase α* gene is associated with fatty

acid composition in a Duroc commercial line. Anim Genet. 2009;40:410–7.

54. Cánovas A, Quintanilla R, Amills M, Pena RN. Muscle transcriptomic profiles in

pigs with divergent phenotypes for fatness traits. BMC Genomics. 2010;11:372.

55. Quintanilla R, Pena RN, Gallardo D, Cánovas A, Ramírez O, Díaz I, et al. Porcine

intramuscular fat content and composition are regulated by quantitative trait loci with

muscle-specific effects. J Anim Sci. 2011;89:2963–71.

56. González-Prendes R, Mármol-Sánchez E, Quintanilla R, Castelló A, Zidi A,

Ramayo-Caldas Y, et al. About the existence of common determinants of gene

expression in the porcine liver and skeletal muscle. BMC Genomics. 2019;20:518.

57. Ramayo-Caldas Y, Mármol-Sánchez E, Ballester M, Sánchez JP, González-Prendes

R, Amills M, et al. Integrating genome-wide co-association and gene expression to

identify putative regulators and predictors of feed efficiency in pigs. Genet Sel Evol.

2019;51:48.

58. Gautier L, Cope L, Bolstad BM, Irizarry RA. affy--analysis of Affymetrix

GeneChip data at the probe level. Bioinformatics. 2004;20:307–15.

59. Smedley D, Haider S, Durinck S, Pandini L, Provero P, Allen J, et al. The BioMart

community portal: an innovative alternative to large, centralized data repositories.

Nucleic Acids Res. 2015;43:W589–98.

60. Cayuela JM, Garrido MD, Bañón SJ, Ros JM. Simultaneous HPLC analysis of α-

tocopherol and cholesterol in fresh pig meat. J Agric Food Chem. 2003;51:1120–4.

61. Purcell S, Neale B, Todd-Brown K, Thomas L, Ferreira MAR, Bender D, et al.

PLINK: a tool set for whole-genome association and population-based linkage analyses.

Am J Hum Genet. 2007;81:559–75.

62. Zhou X, Stephens M. Genome-wide efficient mixed-model analysis for association

studies. Nat Genet. 2012;44:821–4.

63. Karagkouni D, Paraskevopoulou MD, Chatzopoulos S, Vlachos IS, Tastsoglou S,

Kanellos I, et al. DIANA-TarBase v8: A decade-long collection of experimentally

supported miRNA-gene interactions. Nucleic Acids Res. 2018;46:D239–45.

64. Leclercq M, Diallo AB, Blanchette M. Prediction of human miRNA target genes

using computationally reconstructed ancestral mammalian sequences. Nucleic Acids

Res. 2017;45:556–66.

65. Benjamini Y, Hochberg Y. Controlling the False Discovery Rate: A Practical and

Powerful Approach to Multiple Testing. J R Stat Soc Ser B. 1995;57:289–300.

66. Bindea G, Mlecnik B, Hackl H, Charoentong P, Tosolini M, Kirilovsky A, et al.

ClueGO: a Cytoscape plug-in to decipher functionally grouped gene ontology and

pathway annotation networks. Bioinformatics. 2009;25:1091–3.

67. Shannon P, Markiel A, Ozier O, Baliga NS, Wang JT, Ramage D, et al. Cytoscape:

A software environment for integrated models of biomolecular interaction networks.

Genome Res. 2003;13:2498–504.

68. Ballester M, Ramayo-Caldas Y, Revilla M, Corominas J, Castelló A, Estellé J, et al.

Integration of liver gene co-expression networks and eGWAs analyses highlighted

candidate regulators implicated in lipid metabolism in pigs. Sci Rep. 2017;7:46539.

69. Rio DC, Ares M, Hannon GJ, Nilsen TW. Purification of RNA by SDS

solubilization and phenol extraction. Cold Spring Harb Protoc.

2010;2010:pdb.prot5438.

70. Livak KJ, Schmittgen TD. Analysis of relative gene expression data using real-time

quantitative PCR and the $2^{-\Delta\Delta CT}$ method. Methods. 2001;25:402–8.

71. Welch BL. The generalizaton of 'student's' problem when several different

population varlances are involved. Biometrika. 1947;34:28-35.

72. Timoneda O, Balcells I, Córdoba S, Castelló A, Sánchez A. Determination of

Reference microRNAs for relative quantification in porcine tissues. PLoS One.

2012;7:e44413.

73. Paudel Y, Madsen O, Megens H-J, Frantz LAF, Bosse M, Crooijmans RPMA, et al.

Copy number variation in the speciation of pigs: a possible prominent role for olfactory

receptors. BMC Genomics. 2015;16:330.

74. Dueck A, Ziegler C, Eichner A, Berezikov E, Meister G. MicroRNAs associated

with the different human Argonaute proteins. Nucleic Acids Res. 2012;40:9850–62.

75. Groenen MAM, Archibald AL, Uenishi H, Tuggle CK, Takeuchi Y, Rothschild MF,

et al. Analyses of pig genomes provide insight into porcine demography and evolution.

Nature. 2012;491:393–8.

76. Frantz LAF, Schraiber JG, Madsen O, Megens HJ, Bosse M, Paudel Y, et al.

Genome sequencing reveals fine scale diversification and reticulation history during

speciation in Sus. Genome Biol. 2013;14:R107.

77. Steri M, Idda ML, Whalen MB, Orrù V. Genetic variants in mRNA untranslated

regions. Wiley Interdiscip Rev RNA. 2020;9:e1474.

78. Bachtiar M, Jin Y, Wang J, Tan TW, Chong SS, Ban KHK, et al. Architecture of

population-differentiated polymorphisms in the human genome. PLoS One.

2019;14:e0224089.

79. Omariba G, Xu F, Wang M, Li K, Zhou Y, Xiao J. Genome-Wide analysis of

microRNA-related single nucleotide polymorphisms (SNPs) in mouse genome. Sci Rep.

2020;10:5789.

80. Friedman RC, Farh KKH, Burge CB, Bartel DP. Most mammalian mRNAs are

conserved targets of microRNAs. Genome Res. 2009;19:92–105.

81. Gong J, Tong Y, Zhang HM, Wang K, Hu T, Shan G, et al. Genome-wide

identification of SNPs in microRNA genes and the SNP effects on microRNA target

binding and biogenesis. Hum Mutat. 2012;33:254–63.

82. He S, Ou H, Zhao C, Zhang J. Clustering pattern and functional effect of SNPs in

human miRNA seed regions. Int J Genomics. 2018;2018:2456076.

83. Lewis BP, Burge CB, Bartel DP. Conserved seed pairing, often flanked by

adenosines, indicates that thousands of human genes are microRNA targets. Cell.

2005;120:15–20.

84. Bartel DP. MicroRNAs: target recognition and regulatory functions. Cell.

2009;136:215–33.

85. Salomon WE, Jolly SM, Moore MJ, Zamore PD, Serebrov V. Single-molecule

imaging reveals that Argonaute reshapes the binding properties of its nucleic acid

guides. Cell. 2015;162:84–95.

86. Ruda VM, Chandwani R, Sehgal A, Bogorad RL, Akinc A, Charisse K, et al. The

roles of individual mammalian Argonautes in RNA interference in vivo. PLoS One.

2014;9:e101749.

87. Huang V, Zheng J, Qi Z, Wang J, Place RF, Yu J, et al. Ago1 interacts with RNA

polymerase II and binds to the promoters of actively transcribed genes in human cancer

cells. PLoS Genet. 2013;9:e1003821.

88. Ludwig N, Leidinger P, Becker K, Backes C, Fehlmann T, Pallasch C, et al.

Distribution of miRNA expression across human tissues. Nucleic Acids Res.

2016;44:3865–77.

89. Brady MJ, Saltiel AR. The role of protein phosphatase-1 in insulin action. Recent

Prog. Horm. Res. 2001;56:157–73.

90. Schmutz I, Wendt S, Schnell A, Kramer A, Mansuy IM, Albrecht U. Protein

Phosphatase 1 (PP1) is a post-translational regulator of the mammalian circadian clock.

PLoS One. 2011;6:e21325.

91. Liu X, Song B, Li S, Wang N, Yang H. Identification and functional analysis of the

risk microRNAs associated with cerebral low-grade glioma prognosis. Mol Med Rep.

2017;16:1173–9.

92. Wang Y, Viscarra J, Kim SJ, Sul HS. Transcriptional regulation of hepatic

lipogenesis. Nat. Rev. Mol. Cell Biol. 2015;16:678–89.

93. Toole BJ, Cohen PTW. The skeletal muscle-specific glycogen-targeted protein

phosphatase 1 plays a major role in the regulation of glycogen metabolism by

adrenaline in vivo. Cell Signal. 2007;19:1044–55.

94. Ram DR, Ilyukha V, Volkova T, Buzdin A, Tai A, Smirnova I, et al. Balance

between short and long isoforms of cFLIP regulates Fas-mediated apoptosis in vivo.

Proc Natl Acad Sci U S A. 2016;113:1606–11.

95. Mattiotti A, Prakash S, Barnett P, van den Hoff MJB. Follistatin-like 1 in

development and human diseases. Cell. Mol. Life Sci. 2018;75:2339–54.

96. Hargous Y, Hautbergue GM, Tintaru AM, Skrisovska L, Golovanov AP, Stevenin J,

et al. Molecular basis of RNA recognition and TAP binding by the SR proteins SRp20

and 9G8. EMBO J. 2006;25:5126–37.

97. Sebastiani G, Grieco FA, Spagnuolo I, Galleri L, Cataldo D, Dotta F. Increased

expression of microRNA miR-326 in type 1 diabetic patients with ongoing islet

autoimmunity. Diabetes Metab Res Rev. 2011;27:862–6.

98. Kefas B, Comeau L, Erdle N, Montgomery E, Amos S, Purow B. Pyruvate kinase

M2 is a target of the tumor-suppressive microRNA-326 and regulates the survival of

glioma cells. Neuro Oncol. 2010;12:1102–12.

99. Cardoso TF, Coutinho LL, Bruscadin JJ, Diniz WJ da S, Petrini J, Andrade BGN, et

al. Multi-omics approach reveals mir-SNPs affecting muscle fatty acids profile in

Nelore cattle. Genes. 2021;12:67.

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1190 **Table 1:** List of miRNA polymorphisms genotyped in a population of Duroc pigs (N = 345).

microRNA	SSC[a]	Start	End	Strand	SNP	Type	Alt. Allele[b]	Frequency[c]
ssc-miR-339	1	164025972	164026057	-	rs81349391	apical loop	G	0.4138
ssc-miR-130a	2	13296695	13296773	-	rs344472188	apical loop	G	0.1364
ssc-miR-23a	2	65308117	65308186	+	rs333787816	precursor stem	C	0.5398
ssc-miR-30d	4	6948669	6948747	+	rs340704946	precursor stem	G	0.4762
ssc-miR-371	6	56427208	56427285	-	rs320008166	precursor stem	C	0.2955
ssc-miR-429	6	63491921	63492001	+	rs323906663	precursor stem	A	0.2955
ssc-miR-326	9	9581944	9582034	-	rs319154814	apical loop	A	0.5852
ssc-miR-34c	9	39280278	39280357	+	rs321151601	mature region	A	0.0805
ssc-miR-378-2	12	36947443	36947510	+	rs341950320	precursor stem	A	0.1176
ssc-miR-15b	13	100083172	100083269	+	rs334680106	precursor stem	T	0.2706
ssc-miR-486	17	10758818	10758899	-	rs335924546	precursor stem	T	0.3391
ssc-miR-335	18	18341568	18341659	-	rs334590580	precursor stem	C	0.1875

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1192 [a]SSC: porcine chromosome; [b]Alt. Allele: alternative allele of the genotyped SNP; [c]Frequency: alternative allele frequency in the genotyped

1193 population of 345 Duroc pigs.

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Table 2: Number of significant associations between the 12 genotyped miRNA SNPs and the whole set of mRNA expressed in the *gluteus medius* (GM) skeletal muscle (N = 89) and liver (N = 87) tissue samples from Duroc pigs.

SNP	Type	Significant Associations			
		GM (*P*-value < 0.05)	LIVER (*P*-value < 0.05)	GM (*q*-value < 0.1)	LIVER (*q*-value < 0.1)
rs81349391 (1:164026014)	ssc-miR-339 apical loop (A/G)	456	356	0	0
rs344472188 (2:13296736)	ssc-miR-130a apical loop (A/G)	320	1252	0	1
rs333787816 (2:65308181)	ssc-miR-23a precursor stem (T/C)	2075	751	6	1
rs340704946 (4:6948743)	ssc-miR-30d precursor stem (A/G)	563	398	0	0
rs320008166 (6:56427227)	ssc-miR-371 precursor stem (T/C)	270	365	2	8
rs323906663 (6:63491948)	ssc-miR-429 precursor stem (G/A)	328	406	2	8
rs319154814 (9:9581989)	ssc-miR-326 apical loop (G/A)	372	2400	0	0
rs321151601 (9:39280312)	ssc-miR-34c mature region (C/A)	947	392	0	0
rs341950320 (12:36947510)	ssc-miR-378-2 precursor stem (G/A)	792	322	0	0
rs334680106 (13:100083229)	ssc-miR-15b precursor stem (C/T)	277	205	0	0
rs335924546 (17:10758828)	ssc-miR-486 precursor stem (C/T)	1232	171	0	0
rs334590580 (18:18341582)	ssc-miR-335 precursor stem (T/C)	665	219	1	0

1202 **Table 3:** Top significant associations after correction for multiple testing (*q*-value < 0.1) between miRNA SNPs and the mRNA expression of

1203 their potential targets in the *gluteus medius* (GM) skeletal muscle (N = 89) and liver (N = 87) tissue samples from Duroc pigs.

SNP	Type	Tissue	Probe	ID	Gene	δ^a	se^b	*P*-value	*q*-value[c]
rs333787816 (2:65308181)	ssc-miR-23a precursor stem (T/C)	GM	Ssc.1790.1.S1_at	ENSSSCG00000000019	*NUP50*	-0.1052	0.0288	3.849E-04	6.157E-02
			Ssc.12493.1.A1_at	ENSSSCG00000024027	*PAFAH1B2*	-0.2021	0.0565	3.921E-04	6.157E-02
			Ssc.13786.1.A1_at	ENSSSCG00000003857	*ZYG11B*	0.2614	0.0730	4.341E-04	6.157E-02
			Ssc.8682.2.A1_at	ENSSSCG00000014240	*CSNK1G3*	-0.1532	0.0434	4.515E-04	6.157E-02
			Ssc.23976.1.S1_at	ENSSSCG00000012252	*DDX3X*	-0.4318	0.1217	5.095E-04	6.157E-02
			Ssc.4948.1.S1_at	ENSSSCG00000034725	*UBE2R2*	0.1726	0.0493	5.906E-04	6.157E-02
			Ssc.21303.1.S1_at	ENSSSCG00000003630	*AGO1*	0.1016	0.0275	6.775E-04	6.157E-02
			Ssc.24035.2.A1_at	ENSSSCG00000005935	*AGO2*	0.2896	0.0864	8.339E-04	6.157E-02
			Ssc.3706.1.S2_at	ENSSSCG00000040317	*SOD2*	-0.2987	0.0896	8.880E-04	6.157E-02
			Ssc.3706.1.S1_at	ENSSSCG00000040317	*SOD2*	-0.4224	0.1261	8.884E-04	6.157E-02
rs319154814 (9:9581989)	ssc-miR-326 apical loop (G/A)	LIVER	Ssc.9544.2.S1_a_at	ENSSSCG00000016101	*CFLAR*	0.3246	0.0997	1.264E-03	5.785E-02
			Ssc.11661.2.S1_at	ENSSSCG00000009828	*PPP1CC*	0.4508	0.1436	1.784E-03	5.785E-02
			Ssc.31172.3.S1_at	ENSSSCG00000008601	*SDC1*	0.2507	0.0808	2.486E-03	5.785E-02
			Ssc.11044.1.A1_at	ENSSSCG00000003643	*SF3A3*	0.1808	0.0561	2.598E-03	5.785E-02
			Ssc.23242.1.A1_at	ENSSSCG00000039426	*FSTL1*	0.2407	0.078	2.606E-03	5.785E-02
			Ssc.12222.1.S1_at	ENSSSCG00000013233	*CELF1*	-0.1714	0.0574	3.210E-03	5.939E-02
			Ssc.10946.2.A1_at	ENSSSCG00000011917	*NAA50*	0.2853	0.0993	4.070E-03	6.454E-02
			Ssc.4516.2.S1_at	ENSSSCG00000013592	*ELAVL1*	0.1666	0.0598	6.226E-03	8.639E-02

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1205 [a]δ: estimated allele substitution effect; [b]se: standard error of the substitution effect; [c]*q*-value: *q*-value calculated with the false discovery rate

1206 (FDR) approach [63].

Table 4: Significant associations at the nominal level (*P*-value < 0.01) and after multiple testing correction (*q*-value < 0.1; in bold) between 12 genotyped miRNA SNPs and lipid phenotypes recorded in a population of Duroc pigs (N = 345).

SNP	Type	Trait	δ	se	*P*-value	*q*-value
rs81349391 (1:164026014)	ssc-miR-339 apical loop (A/G)	GM (C16:0)	-0.3234	0.1396	2.005E-02	3.772E-01
		GM SFA	-0.4394	0.2030	2.969E-02	3.772E-01
		GM UFA	0.4392	0.2030	2.978E-02	3.772E-01
rs344472188 (2:13296736)	ssc-miR-130a apical loop (T/C)	BFTLR	-1.2084	0.5914	4.087E-02	9.683E-01
rs340704946 (4:6948743)	ssc-miR-30d precursor stem (A/G)	GM (C20:0)	-0.0219	0.0098	2.541E-02	5.779E-01
rs320008166 (6:56427227)	ssc-miR-371 precursor stem (T/C)	GM IMF	0.5368	0.2461	3.104E-02	9.060E-01
rs323906663 (6:63491948)	ssc-miR-429 precursor stem (G/A)	GM IMF	0.5682	0.2482	2.348E-02	8.923E-01
rs319154814 (9:9581989)	**ssc-miR-326 apical loop (G/A)**	**LD (C14:0)**	**0.0885**	**0.0213**	**8.722E-05**	**3.314E-03**
		GM (C14:0)	**0.0700**	**0.0184**	**3.942E-04**	**7.489E-03**
		LD PUFA/MUFA	**-0.0506**	**0.0188**	**7.109E-03**	**7.653E-02**
		LD (C20:1)	**0.0349**	**0.0125**	**8.055E-03**	**7.653E-02**
		LD (C20:4)	-0.2918	0.1268	2.092E-02	1.438E-01
		LD MUFA	0.9512	0.3971	2.485E-02	1.438E-01
		LD (C18:1)	0.8349	0.3624	2.814E-02	1.438E-01
		LD PUFA	-1.0670	0.5004	3.222E-02	1.438E-01
		LD (C18:2)	-0.7266	0.3443	3.405E-02	1.438E-01
		GM (C18:0)	-0.2245	0.1115	4.423E-02	1.681E-01
rs321151601 (9:39280312)	ssc-miR-34c mature region (C/A)	LD (C20:0)	-0.0262	0.0119	2.655E-02	8.171E-01
rs335924546 (17:10758828)	ssc-miR-486 precursor stem (C/T)	GM (C20:4)	0.3396	0.1465	2.100E-02	4.555E-01
		LD (C14:0)	-0.0586	0.0256	2.937E-02	4.555E-01
		LD IMF	0.3300	0.1553	4.748E-02	4.555E-01
rs334590580 (18:18341582)	ssc-miR-335 precursor stem (T/C)	GM (C16:0)	-0.4305	0.1558	5.862E-03	1.201E-01
		GM (C20:0)	0.0331	0.0121	6.321E-03	1.201E-01
		GM UFA	0.5231	0.2278	2.147E-02	2.045E-01

GM SFA	-0.5229	0.2278	2.153E-02	2.045E-01

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1210 [a]δ: estimated allele substitution effect; [b]se: standard error of the substitution effect; [c]*q*-value: q-value calculated with the false discovery rate

1211 (FDR) approach [63]. GM: *gluteus medius* skeletal muscle; LD: *longissimus dorsi* skeletal muscle; BFTLR: backfat thickness at the last rib;

1212 IMF: intramuscular fat content; SFA: saturated fatty acids content; UFA: unsaturated fatty acids content; PUFA: polyunsaturated fatty acids

1213 content; MUFA: monounsaturated fatty acids content; C14:0: myristic content; C16:0: palmitic content; C18:0: stearic content; C18:1: oleic

1214 content; C18:2: linoleic content; C20:0 arachidic content; C20:1: gadoleic content; C20:4: arachidonic content.

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Figure legends

Fig. 1: Schematic representation of (**A**) precursor miRNA sequence hairpin conformation, with mature miRNAs located in the stems (higlighted in blue) and their corresponding seed regions (in red). (**B**) Canonical miRNA-mRNA interactions where each binding site is conformed by 6-7 Watson-Crick pairings (vertical lines) to the seed region of the mature miRNA (2nd to 8th 5' nucleotides). As reported elsewhere [2,47,48], 7mer-m8 and 8mer binding sites are considered to be the most functionally active canonical miRNA-mRNA interactions.

Fig. 2: Principal component analysis plots based on SNPs mapping to: (**A**) the whole-genome (N = 19,720,314 SNPs), (**B**) miRNA genes (N = 285 SNPs), (**C**) Full 3'UTRs (N = 709,343 SNPs), (**D**) 3'UTR 7mer-m8 sites (N = 107,196 SNPs) and (**E**) 3'UTR 8mer sites (N = 33,511 SNPs), respectively.

Fig. 3: Alternative allele frequency distribution of polymorphisms located at miRNA loci in (**A**) Asian domestic pigs (ADM), (**B**) European domestic (EDM) pigs, (**C**) Asian wild boars (AWB) and (**D**) European wild boars (EWB).

Fig. 4: (**A**) SNP density per 100 bp for each analyzed miRNA region considering the full set of 120 porcine whole-genome sequences (WGS) as well as each WGS from the ADM, EDM, AWB and EWB groups. (**B**) SNP density across mature miRNA regions. 1: anchor (1st 5' end position), 2: seed (2nd to 8th position) and 3: supplemental pairing (13th to 18th position). (**C**) SNP density in miRNA loci and flanking regions considering a window of $\pm$ 1kb divided in 10 upstream and downstream 100 bp bins.

Fig. 5: Hepatic mRNA expression levels of the *CFLAR, PPP1CC, SDC1, SF3A3, FSTL1*, *NAA50* and *ELAVL1* genes according to the genotype of the rs319154814 (G/A)

apical loop SNP in the ssc-miR-326 gene. The number of individuals representing each

genotype were: GG (N = 17), GA (N = 37) and AA (N = 32).

Fig. 6: (**A**) Barplots depicting qPCR $\log_2$ transformed relative quantities (Rq) for

CFLAR, PPP1CC and *SF3A3* mRNA transcripts measured in the liver tissue. The

number of individuals representing each genotype were: GG (N = 5) and AA (N = 5).

(**B**) Barplots depicting qPCR $\log_2$ transformed relative quantities (Rq) for ssc-miR-326

transcripts measured in the liver tissue. The number of individuals representing each

genotype were: GG (N = 9) and AA (N = 9).

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Additional Files legends

Additional file 1: Table S1: List of whole-genome sequences from European domestic

pigs (EDM, N = 40), Asian domestic pigs (ADM, N = 40), European wild boars (EWB,

N = 20) and Asian wild boars (AWB, N = 20).

Additional file 2: Table S2: Single nucleotide polymorphisms located in microRNA

genes and their frequencies in European (E) and Asian (A) domestic pigs (DM) and

wild boars (WB).

Additional file 3: Table S3: Means and standard deviations (SD) of fatness and

intramuscular fat content and composition traits recorded in the *gluteus medius* (GM)

and *longissimus dorsi* (LD) muscles of 345 Duroc pigs.

Additional file 4: Fig. S1: Sequence homology between 12 porcine miRNAs harboring

SNPs that have been genotyped in the Lipgen population (N = 345) and their

corresponding human miRNA orthologous sequences.

Additional file 5: Table S4: Set of primers used for the qPCR quantification of three

mRNAs putatively targeted by ssc-miR-326.

Additional file 6: Table S5: Distribution of SNPs mapping to specific genomic

annotated regions (i.e., protein coding genes, exons, introns, 3'UTRs, miRNAs,

lncRNAs, snRNAs, snoRNAs and pseudogenes) and segregating in the set of whole-

genome sequences from 120 Asian and European domestic pigs and wild boars.

Additional file 7: Table S6: List of porcine genes showing 3 or more multi-allelic

SNPs in the set of 120 whole-genome sequences from Asian and European domestic

pigs and wild boars.

Additional file 8: Fig. S2: Number of SNPs present in each of the annotated porcine

miRNA loci that display at least 1 SNP.

Additional file 9: Fig. S3: Venn Diagrams illustrating the sharing of SNPs located at

the (**A**) *precursor*, (**B**) *mature* and (**C**) *seed* regions of miRNA loci amongst Asian

domestic pigs (ADM), Asian wild boars (AWB), European domestic pigs (EDM) and

European wild boars (EWB).

Additional file 10: Table S7: Description of SNPs mapping to porcine mature miRNA

loci (N = 409) and segregating in the set of 120 whole-genome sequences from Asian

and European domestic pigs and wild boars.

Additional file 11: Table S8: Whole-genome sequencing statistics for the five Duroc

boars that sired the Lipgen population (N = 345).

Additional file 12: Fig. S4: Boxplot distribution depicting whole-genome sequencing

statistics for the five Duroc boars that sired the Lipgen population (N = 345).

Additional file 13: Table S9: List of SNPs mapping to 370 porcine miRNA loci and

segregating in the 5 Duroc boars that sired the Lipgen population (N = 345).

Additional file 14: Table S10: Association analyses between SNPs in miRNA genes

and the mRNA levels of their predicted targets in the *gluteus medius* (GM) skeletal

muscle and liver tissues of Lipgen pigs.

Additional file 15: Table S11: Genomic coordinates and sequence of the seed

(miRNA) and its predicted 3'UTR binding site (mRNA) for putative miRNA-mRNA

pairs showing significant associations after multiple testing correction (q-value < 0.1).

Additional file 16: Table S12: Association analyses between miRNA SNPs and the

mRNA expression profiles of the whole set of expressed probes/genes in the porcine

gluteus medius (GM) skeletal muscle and liver tissues from Lipgen pigs.

Additional file 17: Table S13: Pathway enrichment analyses of the lists of

probes/genes significantly associated at the nominal level (P-value < 0.05) with miRNA

SNP genotypes.

Additional file 18: Table S14: Association analyses between miRNA SNPs and fatness

and intramuscular fat content and composition traits recorded in the *gluteus medius*

(GM) and *longissimus dorsi* (LD) skeletal muscles of 345 Duroc pigs.

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1304 **Fig. 1:**

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Fig. 2:

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1354   **Fig. 6:**

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Variability in porcine microRNA genes and its association with lipid composition and mRNA expression phenotypes

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Additional file 4: Fig. S1: Sequence homology between 12 porcine miRNAs harboring SNPs that have been genotyped in the Lipgen population (N = 345) and their corresponding human miRNA orthologous sequences.

hsa-miR-339-5p UCCCUGUCCUCCAGGAGCUCACG	hsa-miR-326 CCUCUGGGCCCUUCCUCCAG
ssc-miR-339 UCCCUGUCCUCCAGGAGCUCA	ssc-miR-326 CCUCUGGGCCCUUCCUCCAG
hsa-miR-130a-3p CAGUGCAAUGUUAAAAGGGCAU	hsa-miR-34c-5p AGGCAGUGUAGUUAGCUGAUUGC
ssc-miR-130a CAGUGCAAUGUUAAAAGGGCAU	ssc-miR-34c AGGCAGUGUAGUUAGCUGAUUGC
hsa-miR-23a-3p AUCACAUUGCCAGGGAUUUCC	hsa-miR-378a-3p ACUGGACUUGGAGUCAGAAGGC
ssc-miR-23a AUCACAUUGCCAGGGAUUUCC	ssc-miR-378-2 ACUGGACUUGGAGUCAGAAGGC
hsa-miR-30d-5p UGUAAACAUCCCCGACUGGAAG	hsa-miR-15b-5p UAGCAGCACAUCAUGGUUUACA
ssc-miR-30d UGUAAACAUCCCCGACUGGAAG	ssc-miR-15b UAGCAGCACAUCAUGGUUUACA
hsa-miR-371a-5p ACUCAAACUGUGGGGGCACU	hsa-miR-486-5p UCCUGUACUGAGCUGCCCCGAG
ssc-miR-371-5p ACUCAAACUGUGGGGGCACUUU	ssc-miR-486 UCCUGUACUGAGCUGCCCCGAG
hsa-miR-429 UAAUACUGUCUGGUAAAACCGU	hsa-miR-335-5p UCAAGAGCAAUAACGAAAAAUG
ssc-miR-429 UAAUACUGUCUGGUAAUGCCGU	ssc-miR-335 UCAAGAGCAAUAACGAAAAAUG

Additional file 8: Fig. S2: Number of SNPs present in each of the annotated porcine

miRNA loci that display at least 1 SNP.

Additional file 9: Fig. S3: Venn Diagrams illustrating the sharing of SNPs located at the (**A**) *precursor*, (**B**) *mature* and (**C**) *seed* regions of miRNA loci amongst Asian domestic pigs (ADM), Asian wild boars (AWB), European domestic pigs (EDM) and European wild boars (EWB).

Additional file 12: Fig. S4: Boxplot distribution depicting whole-genome sequencing statistics for the five Duroc boars that sired the Lipgen population (N = 345).