**Genome-Wide Identification of RNA Editing Sites Affecting Intramuscular Fat in Pigs**

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**Simple Summary:** In order to investigate RNA editing sites affecting IMF (which is associated with pork quality and human insulin resistance.), we analyzed the transcriptome and genome sequencing data of a high- and low- groups composed of full-sib pairs pig with opposite IMF phenotypes. Finally, a total of 36 nonredundant RNA editing sites in the longissimus dorsi muscle, which may reveal the potential importance of RNA editing in IMF were identified. Four were selected as candidate sites associated with IMF. Our findings provide some new insights of RNA editing function in pig longissimus dorsi muscle.

**Abstract:** Intramuscular fat (IMF) is essential for improving the palatability and flavor of meat, and it is strongly associated with human insulin resistance. RNA editing is a widespread regulating event in different tissues. Here, we investigated the global RNA editing difference of two groups of pig with different IMF contents to find the potential editing sites affecting IMF. In this research, RES-Scanner and REDItools were used to identify RNA editing sites based on the whole genome and transcriptome sequencing data of the high and low groups composed of three full-sib pairs with opposite IMF phenotypes. A total of 295 RNA editing sites were investigated in the longissimus dorsi muscle, and 90.17% of these sites caused A to G conversion. After annotation, most editing sites were located in noncoding regions (including five sites located on the 3′ UTR regions). Five editing sites (including two sites that could lead to nonsynonymous amino acid changes) were located in the exons of genes. A total of 36 intergroup (high and low IMF) differential RNA editing sites were found in 33 genes. Some candidate editing sites, such as sites in acyl-coenzymeA: cholesterol acyltransferase 1 (ACAT1), coatomer protein, subunit alpha (COPA), and nuclear receptor coactivator 3 (NCOA3), were selected as candidate RNA editing sites associated with IMF. One site located on the 3′ UTR region of growth hormone secretagogue receptor (GHSR) may regulate GHSR expression by affecting the interaction of miRNA and mRNA. In conclusion, we identified a total of 36 nonredundant RNA editing sites in the longissimus dorsi muscle, which may reveal the potential importance of RNA editing in IMF. Four were selected as candidate sites associated with IMF. Our findings provide some new insights of RNA editing function in pig longissimus dorsi muscle which useful for pig IMF breeding or human insulin resistances research.

**Keywords:** high-throughput sequencing; RNA editing; muscle growth; intramuscular fat; swine

1. Introduction

RNA editing is the post-transcriptional or co-transcriptional process which could make transcripts more complicated and results in potential functional consequences [1,2]. RNA editing sites have
been found in different vertebrates, including humans [3–6], mice [4], pigs [1,7,8], bovines [9], and chickens [10]. The RNA editing events were usually catalyzed by adenosine deaminase RNA specific (ADAR), and represent the form of A-to-I in mammals [11,12]. RNA editing sites in coding regions of genes can alter the amino acid sequences of proteins [7], and the RNA editing sites in non-coding regions such as 3’ UTR are involved in editing of miRNA seed regions [13] and target sequences in mRNA [14]. Moreover, RNA editing sites in intergenic region may associated with nuclear retention [15], and in introns may disrupt alternative splicing [16]. In humans, RNA editing is associated with disease [3–5]; in animals, RNA editing was found to be associated with embryo and tissue development [1], production traits [17], and Newcastle disease virus [18].

Intramuscular fat (IMF) is highly related to insulin resistance [19] and is highly desirable for meat quality factors such as meat flavor, tenderness, and palatability [20]. Therefore, studies on the mechanisms of IMF development are important for pork quality and human diseases such as type 2 diabetes, obesity, and so on. In pigs, many regulators, such as melatonin [21]; mesenteric estrogen- dependent adipogenesis gene (Medag) [22]; IncRNAs, such as MSTRG.604206 (3459 bp to fatty acid synthase (FASN)) and MSTRG.426159 (459614 bp to LOC102160997) [23]; miR-130a [24]; and miR-125a-5p [25], are significantly related to pig IMF content. These regulatory factors mostly focus on small RNAs, IncRNAs, and protein-coding genes. Studies on the effects of RNA editing sites in porcine IMF development are lacking.

In order to mining the potential effective RNA editing sites on porcine IMF, we first investigate RNA editing sites based on strand-specific RNA sequencing and whole-genome resequencing data of three full-sib paired 240-day-old pigs with different IMF contents. And then, we predicted the potential function of RNA editing sites by binding energy analysis and protein interaction network analysis. And finally, several candidate editing site affecting IMF content. This study not only explores the different RNA editing events between pigs with different IMF contents, but also provides new insight into the functional analysis of RNA editing sites.

2. Materials and Methods

2.1. Ethics Statements

All animals experiment in our research were carried out under the ethics approval (No. IASCAAS-AE-09) of Animal Ethics Committee of the Institute of Animal Science, Chinese Academy of Agricultural Sciences on 7 September 2018.

2.2. Sample Collection and Nucleic Acid Isolation

The longissimus dorsi muscle (10th–11th rib, muscle) were sampling from six 240-day-old Large White X Min pig F2 individuals (3 pairs of full-sib, Table 1). All the tissues were snap frozen in liquid nitrogen and stored in −80 °C refrigerator until use. Total genomic DNA was isolated from the muscle tissues using phenol–chloroform protocol. Total RNA was extracted using TRlzol reagent (Invitrogen, Carlsbad, CA, USA) according to the manufacturer’s instructions. The quantification and quality standards of DNA samples were: (1) optical density (OD) 260/280 ratios of 1.8–2.0, (2) total contents >3 µg. The quality standard of RNA samples mainly was RNA integrity number (RIN) scores ≥7.

| ID    | Group | Group ID | Father ID | Mother ID | IMF a Content (%) |
|-------|-------|----------|-----------|-----------|-------------------|
| 19805 | L     | L1       | 721205    | 723604    | 0.9               |
| 1015105 | L     | L2       | 706601    | 706204    | 1.41              |
| 1119609 | L     | L3       | 700105    | 709602    | 1.08              |
| 19809  | H     | H1       | 721205    | 723604    | 5.56              |
| 1015103 | H     | H2       | 706601    | 706204    | 5.94              |
| 1119605 | H     | H3       | 700105    | 709602    | 7.51              |

* IMF: Intramuscular fat.
2.3. Strand-Specific Transcriptome and Whole-Genome Sequencing

Muscle DNA and RNA sequencing libraries were generated using oligo (dT) beads by NEBNext® Ultra™ Directional RNA Library Prep Kit for Illumina® (NEB, Ipswich, MA, USA) and TrSseq Nano DNA HT Sample Preparation Kit (Illumina, San Diego, CA, USA) according to the manufacturer’s recommendations. Then, the sequencing libraries were monitoring by real-time PCR on an Agilent Bioanalyzer 2100 system (Agilent, Lexington, MA, USA) for quantification and quality assessment. At last, libraries constructed above were sequenced on Illumina HiSeq (Illumina, San Diego, CA, USA) platform in Novogene Bioinformatics Technology Cooperation (Beijing, China) by 150 bp paired-end sequencing strategy. All of the clean data have been submitted to the Genome Sequence Archive, with the accession number CRA001645.

2.4. RNA Editing Investigation

In order to strengthen robustness of detection, the RNA editing sites were detected using RES-Scanner with the main parameters —ss 1 —uniqTag 1 and default thresholds in user manual [26]. And then, the REDItoolBlatCorrection.py script of REDItools was used for correction [27]. The workflow of RNA editing sites detection were shown as Figure 1. The reference genome, SNP database, and genomic feature files were all download from Ensembl (ftp://ftp.ensembl.org/pub/release-98/fasta/sus_scrofa/). In order to decrease the false positive rate, strict standards were used for RNA editing sites detection as follow: (a) DNA counts ≥10, and homozygous >0.95 (b) counts of RNA reads which differ from DNA genotype ≥3, (c) RNA editing level (editing counts/ total reads of this locus) ≥0.05 and false discovery rate (FDR) <0.05, and (d) candidate editing site is not located within regions of ≥3 residues and ≤6 intronic bases of a splice site.

Figure 1. Overview of the workflow of RNA editing sites detection.
2.5. Validation of Candidate RNA Editing Sites

All of the 295 sequences (50 bp upstream and 50 downstream of each candidate RNA editing sites) were retrieved from the reference genomes. After editing sites altering, all of the sequences were queried against the standard database of expressed sequences tags (ESTs) of swine in NCBI using BLAST algorithm. Five editing sites which could not be found in the DREP (Database of RNA Editing in Pig, http://www.manet.org/editing/main.php) and ESTs databases were randomly selected to validate the prediction accuracy of RES-Scanner. DNA and RNA were isolated from the muscle tissues of the same individuals for sequencing. The primers (Table S1) for sanger sequencing were designed using the software of primer6 and synthesized by Invitrogen Inc. (Shanghai, China). The sanger sequencing was carried by Liuhetong Inc. (Beijing, China). The sites were considered verified if the cDNA sequence was heterozygous while the corresponding DNA sequencing was homozygous.

2.6. Annotation of RNA Editing Sites

Annotate variation (ANNOVAR) [28] was used to annotate the genomic features and covered (or nearby) genes of RNA editing sites. Seven types of genomic features which were exonic, ncRNA, downstream, intergenic, UTR, and intronic, and gene Ensembl ID were retrieved. To determine whether the RNA editing sites located in repetitive elements, we used AnnotateTable.py script of REDItools [27].

2.7. Functional Enrichment Analysis

We used bioDBnet [29] to convert the Ensembl ID to gene symbol and used Metascape (metascape.org) [30] to perform a functional enrichment analysis based on Gene Ontology (GO) biological processes and Kyoto Encyclopedia of Genes and Genomes (KEGG) pathway terms. GO/KEGG terms with the threshold of FDR < 0.05 were considered significantly enriched. For each given gene list, protein-protein interaction enrichment analysis has been carried out with the following databases: BioGrid [31], InWeb_IM [32], OmniPath [33]. The resultant network contains the subset of proteins that form physical interactions with at least one other member in the list. If the network contains between 3 and 500 proteins, the Molecular Complex Detection (MCODE) algorithm [34] has been applied to identify densely connected network components.

2.8. Trait Differential RNA Editing Sites Investigation

As described by Peng et al. [35], the sites only edited in high IMF individuals (more than two individuals with editing level > 0.15) but not in low IMF individuals were defined as hyper-editing sites. Conversely, the sites only edited in low IMF individuals (more than two individuals with editing level > 0.15) but not in high IMF individuals were defined as hypo-editing sites.

2.9. Impacts of RNA Editing Events on miRNA-mRNA Interactions

For each hyper- and hypo-editing site located at 3’UTR, we extracted the editing site and the upstream and downstream 25 base pairs (total of 51 bp) from reference genome as an unedited-type (UT) sequence and changed the editing site from A-to-G as an edited-type (ET) sequence. Then, miRANDA [36] was used to calculated the binding energy (with the threshold of −7 kCal/Mol) and to predict the miRNA target sites on UT and ET sequences. The miRNA-mRNA interactions not in UT, but in ET, were defined as interaction losses. The miRNA-mRNA interactions not in ET, but in UT, were defined as interaction gain. The change of binding energy > 2 kCal/Mol, were defined as change.

2.10. Impacts of RNA Editing Sites on Protein Function

For the A-to-G editing site on the exons, SIFT software [37] was used to predict the on-protein function. Variants with SIFT scores ranged from 0.0 to 0.05 were considered deleterious.
3. Results

3.1. The Landscape of RNA Editing Sites in Swine Longissimus Dorsi Muscle

To investigate the RNA editing sites at the transcriptome-wide level in longissimus dorsi muscle, DNA sequencing and matched strand-specific RNA sequencing were performed for the six pigs. After quantification and quality filtering, a total of 542 million RNA reads were acquired from the six pigs, with an average mapping rate of 76.7%. A summary of the high-throughput sequencing result is provided in Table 2. After the strand-specific RNA editing analysis, a number of 295 sites (including 266 A-to-G editing sites) were detected (Figure 2A).

| Individual | Total Reads | Mapped Rate | Total Reads | Mapping Rate | Coverage |
|------------|-------------|-------------|-------------|--------------|----------|
| H1         | 86,985,200  | 77.1%       | 423,325,180 | 87.2%        | 81.3%    |
| H2         | 80,100,158  | 77.7%       | 476,288,926 | 90.4%        | 84.3%    |
| H3         | 81,579,006  | 75.3%       | 435,526,280 | 88.9%        | 82.1%    |
| L1         | 115,906,328 | 75.8%       | 500,321,520 | 87.2%        | 85.9%    |
| L2         | 91,861,266  | 77.3%       | 476,238,994 | 89.4%        | 84.2%    |
| L3         | 86,400,546  | 77.25%      | 456,625,288 | 86.8%        | 81.9%    |

The coverage was estimated based autosomal and X chromosomes.

Consistent with the previous study in pigs, RNA editing levels of the most editing sites in muscle were low [1,3,8]. Most of the editing sites were under 10.0%, with an average editing level of 21.06% (Figure 2B). Of all detected editing sites 67.46% were located in retained introns (Figure 2C). Also consistent with previous reports, the editing sites were most located in the intronic regions of genes, and these suggested that the RNA editing sites may impact splicing [16]. Only five editing
sites were located in the exons, and two of them could cause protein coding changing (Figure 2D). The A-to-G changing of the sites locate on ssc4:14706035 will lead to the glutamine-to-arginine changing of the coding protein of ENSSSCG00000043481, and the sites located on ssc8:84574384 will cause the alanine-to-valine changing of ubiquitin-specific peptidase 38 (USP38). In the results of SIFT prediction, the editing of the sites in USP38 was deleterious to its protein function ($p < 0.05$). As synonymous variants can disrupt transcription, splicing and mRNA stability [38], we used mfold [39] to predict the stability of the RNA structure of the genes containing the synonymous editing sites. Unfortunately, the three synonymous RNA editing sites could not cause any stability changes. Moreover, there are no sites were found to be located in the exon start_0base of the alternative splicing (AS) appearing in the six pigs longissimus dorsi muscle.

3.2. Analysis of the Editing Sites Distribution in Chromosome and Repetitive Regions

According to the results of the genomic distribution of the editing sites, the average distances between neighboring editing sites in the longissimus dorsi muscle did not have significant difference across chromosomes ($p < 0.05$). And, this indicated that editing events in the longissimus dorsi muscle have little chromosomal bias (Figure 3A). Consistent with previous studies [1,3,8], 85.76% of the repetitive editing sites were located in short interspersed nuclear elements (SINEs) (Figure 3B). In contrary, only 1.69% of the repetitive editing sites located in long interspersed nuclear elements (LINEs). And included in the SINES, most of the editing sites were located in Pre0_SS elements, with the following PRE1e, PRE1f2, PRE1f, PRE1g, PRE1h, PRE1d, PRE1i, SINE1A_SS, MIR, PRE1c, SINE1_SS, and SINE1D_SS (Figure 3C).

**Figure 3.** Distribution of RNA editing sites across the genome. (A) Distribution of RNA editing sites across chromosomes. The peaks showed the appearance of editing, and the valleys showed the disappearance of editing sites. (B) Distribution of RNA editing sites across repetitive classes. SINEs, short interspersed nuclear elements; LINEs, long interspersed nuclear elements. (C) Distribution of RNA editing sites across repetitive elements.
3.3. Validation of Candidate RNA Editing Events

According to the in-silico blasting on NCBI, a total number of 235 (79.66%) editing sites were found covering by more than one EST clone (Table S2), and a total number of 206 (69.83%) editing sites were found in more than one edited EST sequence (Figure 4A). When searching in the database of RNA editing in pig (DREP, http://www.rnanet.org/editing/main.php) [8], about 69.17% (184/266) A-to-G RNA editing sites were found previously reported (Figure 3A). We selected five predicted editing sites that were included neither in the DREP database nor in EST database (totally 35 sites) to perform sanger sequencing. In the results of sanger sequencing, three of the five selected sites could be validated (Figure 4B and Table S1). In all, maybe 21 sites (35 × 0.6) investigate in our research were first investigated in pig longissimus dorsi muscle.

![Figure 4](image)

Figure 4. Validation of RNA editing sites. (A) Venn diagram of the RNA editing sites searching in other databases. EST, Expression site tags; DREP, Database of RNA Editing in Pig; ALL, all of the 295 editing sites detected in this research; ESTSITE, EST clones with known editing sites. (B) Sanger sequencing plot showing different nucleotide (A and G) between genomic DNA and RNA at same position (editing site at Sus scrofa chromosome 13: 111023145-). The sites are highlighted in red lines.

3.4. Functional Enrichment and Interactome Analysis of RNA Editing Sites

After annotation, the RNA editing sites were located in or nearby 254 genes. We converted the ensembl number of these genes to official gene symbols, and then submitted genes without synonymous editing sites to Metascape for GO and interactome analysis. Together, we found that genes with editing sites participate in the function of intermuscular fat development, such as MAPK6/MAPK4 signaling. And most of genes with editing sites were associated with muscle development such as myeloid cell development, adherens junction, and skeletal system development. Moreover, genes with editing sites participate in some other biological functions such as the vascular endothelial growth factor receptor signaling pathway, the response to bronchodilators, etc. (Figure 5A). All the genes could interact with each other in four networks, including one large network and three small networks (Figure 5B).
3.5. Investigation of Differential RNA Editing Sites in Low and High—IMF Pig Longissimus Dorsi Muscle

A total of 36 hyper- or hypo-editing sites in 33 genes were investigated in all of the editing sites. Most of the sites were also in the introns of genes, and only two sites were located in the 3’UTR region of the genes (Figure 6A). Among the 36 differential RNA editing sites, 31 were hyper-editing sites and only 5 were hypo-editing sites in high-IMF individuals (Figure 6B). Most of the genes with differential editing sites participate in the function of muscle or intermuscular fat development, such as endomembrane system organization, adherens junction, muscle contraction, and the transmembrane receptor protein tyrosine kinase signaling pathway (Figure 6C). Most of the genes are involved in muscle contraction, muscle system process, regulation of muscle system process, etc. (Figure 6D and Table S3). These genes are located in an independent cluster that contains acyl-coenzymeA:cholesterol acyltransferase 1 (ACAT1), solute carrier family 8 member A1 (SLC8A1), phosphodiesterase
3A (PDE3A), catenin alpha 3 (CTNNA3), tripartite motif containing 72 (TRIM72), and follistatin like 1 (FSTL1), with the central of growth hormone secretagogue receptor (GHSR).

Figure 6. Distribution and function analysis of genes containing differential RNA editing sites. (A) Genomic features of hyper- and hypo-editing sites in genes. (B) Cluster heatmap of differential editing sites. Different color represents different editing level. (C) Most significant GO biological function of the genes with hyper- or hypo-editing sites. (D) The interactome network of the genes with hyper- or hypo-editing sites. Different biological functions are represented in different colors.

3.6. Impacts of RNA Editing Events on Protein Function

The protein–protein interaction (PPI) networks were also constructed using all of the genes containing edited sites in the longissimus dorsi muscle (Figure 7A). As shown in the four MCODE component networks containing the densely-connected proteins, three proteins coded by genes containing differential editing sites could interact with other proteins (Figure 7B) and these three proteins were coatomer protein, subunit alpha (COPA), ACAT1, and nuclear receptor coactivator 3 (NCOA3).
3.7. Impacts of RNA Editing Sites on miRNA–mRNA Interactions

When we used miRANDA to calculate the binding ability of miRNA target sites on UT and ET sequences, a total of 4 miRNA–mRNA interaction losses and 2 miRNA-mRNA interaction gains were found in the 3′ UTR region of zinc finger protein 543 (ZNF543). Five miRNA-mRNA interaction changes were found in the 3′ UTR region of GHSR. Seven miRNA-mRNA interaction gains were found in the 3′ UTR region of GHSR (Table 3).

Table 3. miRNAs 1 that represent differential binding ability to different type of editing sites.

| Gene   | miRNA ID   | Type   |
|--------|------------|--------|
|        | ssc-miR-15b | change |
|        | ssc-miR-20a-3p | gain   |
|        | ssc-miR-216 | gain   |
|        | ssc-miR-217 | gain   |
|        | ssc-miR-103 | change |
|        | ssc-miR-107 | change |
|        | ssc-miR-16  | gain   |
| GHSR 2 | ssc-miR-221-3p | loss   |
|        | ssc-miR-503 | change |
|        | ssc-miR-497 | gain   |
|        | ssc-miR-222 | loss   |
|        | ssc-miR-4339 | change   |
|        | ssc-miR-187 | gain   |
|        | ssc-miR-2483 | gain   |
|        | ssc-miR-9858-5p | loss   |
| ZNF543 3 | ssc-miR-145-3p | loss   |
|        | ssc-miR-30e-3p | loss   |
|        | ssc-miR-664-3p | loss   |
|        | ssc-miR-9849-5p | gain   |
|        | ssc-miR-9861-5p | loss   |
|        | ssc-miR-10383 | gain   |

1 miRNAs, microRNAs; 2 GHSR, growth hormone secretagogue receptor; 3 ZNF543, Zinc finger protein 543.

4. Discussion

In this study, 11 types of RNA editing were detected, including all possible base substitutions as follows: A-to-G, A-to-C, A-to-T, C-to-A, C-to-T, G-to-A, G-to-C, G-to-T, T-to-A, T-to-C, and T-to-G.
Overall, the A-to-G substitution was the most common, accounting for up to 90.17% of the identified RNA editing sites. Consistent with previous reports, the average editing level of the A-to-G sites was low overall [1,3,8]. Genes containing editing sites may be involved in various biological functions. However, the genes containing different editing levels in different IMF content groups were mainly involved in the biological function related to myocyte development and lipidosis (Figure 6C and Table S3). As we know, endomembrane system organization and adherens junction process are the basic processes of myofibrillogenesis. The transmembrane receptor protein tyrosine kinase signaling pathway could regulate energy metabolism and is negatively correlated with obesity [40].

As most of the genes with differential editing sites are associated with the function of muscle, we inferred that most of the sites may have function. As most of the sites were located in the introns, we investigated the alternative splicing (AS) appearing in the six pigs’ longissimus dorsi muscle, and searched the alternative splicing status of all the DE genes. Finally, 23 genes have alternative splicing (Table S4), but no sites were found to be located in the exon start_0base. And moreover, there are also no sites located on the alternative splicing.

Among the 23 genes, ACAT1 and NCOA3 were found in one PPI MCODE network. In this MCODE network, ACAT1 and NCOA3 could directly interact with Ras Homolog Family Member A (RHOA), and F-Box and WD repeat domain containing 7 (FBXW7) proteins (Figure 6B). FBXW7 was reported to be associated with lipid metabolism [41]. In previous research, ACAT1 was consider as one of the candidate genes affect ribeye area and backfat thickness [42]. And in the research of Gu et al. [43], stabilization of FASN by ACAT1-mediated GNPAT acetylation could promote lipid metabolism. Emerging evidence from previous studies using animal models also suggests that the NCOA3 plays a critical role in lipid metabolism as well as adipogenesis and obesity [44]. So, we and we inferred that editing sites in these two genes may have influences associated with IMF deposition or metabolism.

COPA was also found in another PPI MCODE network containing Chaperonin Containing TCP1 Subunit 6B (CCT6B), Replication Protein A1 (RPA1), DEAD-Box Helicase 6 (DDX6), Proteasome 20S Subunit Beta 2 (PSMB2), DnaJ Heat Shock Protein Family (HSP40) Member A1 (DNAJA1), Cyclin Dependent Kinase 2 (CDK2), and Tubulin Alpha 8 (TUBA8). Overexpression of CCT6B significantly inhibited fibroblast migration and collagen synthesis (all \( p < 0.05 \)) [45]. DDX6 plays essential roles in adipogenesis and in the alternative splicing of peroxisome proliferator activated receptor gamma (PPARG) and Lipin 1 (Lpin1) [46]. PSMB2 can be expressed in human brown adipose tissue [47]. HSP40 plays a vital role to obesity-induced insulin resistance and type 2 diabetes (T2D) [48]. Decreased muscle protein degradation may be regulated though DNAJA1 pathway in swine [49]. m⁶A-dependent CDK2 expression, mediated by FTO Alpha-Ketoglutarate Dependent Dioxygenase (FTO) and YTH N6-Methyladenosine RNA Binding Protein 2 (YTHDF2), contributes to adipogenesis inhibition [50]. In Martinez-Montes [17], the COPA: g.48255C>T polymorphisms showed effects on backfat thickness. As most of the protein in this MCODE network is associated with adipogenesis, we inferred that COPA may be a candidate gene affecting IMF content. And more research, such as RNA pulldown and western blotting experiment could be performed to validate the editing events’ effect on interacted proteins.

As the differences in the expression levels of the transcripts in ZNF543 were minimal between the high- and low-IMF groups, we inferred that although the editing sites in the 3’ UTR region of ZNF543 have miRNA–mRNA interaction gain/loss or change, the role of this interaction may have no effect on ZNF543 expression. There are possible post-translational effects that might justify an effect of ZNF543 on IMF. One transcript, named GHSR-202, showed a 10.59-fold change expression difference between the high- and low-IMF groups. We searched all of the potential interactions with miRNAs in PubMed. ssc-miR-15b, ssc-miR-16, ssc-miR-103, and ssc-miR-221-3p were reported to have different expressions in the myofiber between lean and obese pig breeds [51]. ssc-miR-16 was also reported to be upregulated in the muscle of obese minipigs [52]. ssc-miR-103 and ssc-miR-221-3p are associated with meat quality [53]. Ghrelin exerts direct peripheral effects on lipid metabolism,
including increasing white adipose tissue mass, stimulating lipogenesis in the liver, and taste sensitivity modulation. Ghrelin and its receptor GHSR are involved in obesity [54–56]. Combined with the results of GHSR being the center of the GO cluster, we speculated that GHSR may play a crucial part in IMF regulation. The editing site in the 3′ UTR of GHSR may contribute to the lipogenesis by regulating the miRNA of ssc-miR-15b, ssc-miR-16, ssc-miR-103, or ssc-miR-221-3p. To determine the target miRNA, further research, such as dual luciferase reporter assay should be performed.

5. Conclusions

In this study, a total of 295 RNA editing sites were identified using strand-specific RNA and whole-genome DNA sequencing data in pig longissimus dorsi muscle. The functional analysis of the genes with hyper- and hypo-editing sites revealed the potentially functional importance of RNA editing in porcine IMF regulation. Some candidate editing sites, such as in ACAT1, NCOA3, COPA, and especially GHSR, should be deeply studied in future research. More research such as RNA pulldown, western blotting and dual luciferase reporter assay should be done to explore the relationship between these candidate editing sites and interacted genes and proteins. Our findings provide some new RNA editing sites and some new insights of RNA editing function which useful for pig IMF breeding or human insulin resistances research.

Supplementary Materials: The following are available online at http://www.mdpi.com/2076-2615/10/9/1616/s1, Table S1: Primer and the sequence information of editing sites for validation, Table S2: Blast results of retrieved sequence containing RNA editing sites, Table S3: Cluster of GO containing genes with differential RNA editing sites in different group, Table S4: Alternative splicing type of genes with intronic RNA editing sites.

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References

1. Yang, Y.; Zhu, M.; Fan, X.; Yao, Y.; Yan, J.; Tang, Y.; Liu, S.; Li, K.; Tang, Z. Developmental atlas of the RNA editome in Sus scrofa skeletal muscle. DNA Res. 2019, 26, 261–272. [CrossRef] [PubMed]
2. Bass, B.L. RNA editing by adenosine deaminases that act on RNA. Annu. Rev. Biochem. 2002, 71, 817–846. [CrossRef] [PubMed]
3. Goncharov, A.O.; Kliuchnikova, A.A.; Nasaev, S.S.; Moshkovskii, S.A. RNA Editing by ADAR Adenosine Deaminases: From Molecular Plasticity of Neural Proteins to the Mechanisms of Human Cancer. Biochemistry 2019, 84, 896–904. [CrossRef] [PubMed]
4. Levitsky, L.I.; Kliuchnikova, A.A.; Kuznetsova, K.G.; Karpov, D.S.; Ivanov, M.V.; Pyatnitskiy, M.A.; Kalinina, O.V.; Gorshkov, M.V.; Moshkovskii, S.A. Adenosine-to-Inosine RNA Editing in Mouse and Human Brain Proteomes. Proteomics 2019, 19, 1900195. [CrossRef] [PubMed]
5. Gatsiou, A.; Vlachogiannis, N.; Lunella, F.F.; Sachse, M.; Stellos, K. Adenosine-to-Inosine RNA Editing in Health and Disease. Antioxid. Redox Signal. 2018, 29, 846–863. [CrossRef] [PubMed]
6. Zhou, Z.Y.; Hu, Y.; Li, A.; Li, Y.J.; Zhao, H.; Wang, S.Q.; Otecko, N.O.; Zhang, D.; Wang, J.H.; Liu, Y. Genome wide analyses uncover allele-specific RNA editing in human and mouse. Nucleic Acids Res. 2018, 46, 8888–8897. [CrossRef] [PubMed]
7. Zhang, Y.; Zhang, L.; Yue, J.; Wei, X.; Wang, L.; Liu, X.; Gao, H.; Hou, X.; Zhao, F.; Yan, H. Genome-wide identification of RNA editing in seven porcine tissues by matched DNA and RNA high-throughput sequencing. J. Anim. Sci. Biotechnol. 2019, 10. [CrossRef]
8. Wang, Z.; Feng, X.; Tang, Z.; Li, S.C. Genome-Wide Investigation and Functional Analysis of Sus scrofa RNA Editing Sites across Eleven Tissues. *Genes* **2019**, *10*, 327. [CrossRef]
9. Bakhtiarizadeh, M.R.; Salehi, A.; Rivera, R.M. Genome-wide identification and analysis of A-to-I RNA editing events in bovine by transcriptome sequencing. *PLoS ONE* **2018**, *13*, e0193316. [CrossRef]
10. Shafiei, H.; Bakhtiarizadeh, M.R.; Salehi, A. Large-scale potential RNA editing profiling in different adult chicken tissues. *Anim. Genet.* **2019**, *50*, 460–474. [CrossRef]
11. Tan, M.H.; Li, Q.; Shanmugam, R.; Piskol, R.; Kohler, J.; Young, A.N.; Liu, K.I.; Zhang, R.; Ramaswami, G.; Ariyoshi, K. Dynamic landscape and regulation of RNA editing in mammals. *Nature* **2017**, *550*, 249–254. [CrossRef] [PubMed]
12. Bazak, L.; Haviv, A.; Barak, M.; Jacob-Hirsch, J.; Deng, P.; Zhang, R.; Isaacs, F.J.; Rechavi, G.; Li, J.B.; Eisenberg, E. A-to-I RNA editing occurs at over a hundred million genomic sites, located in a majority of human genes. *Genome Res.* **2014**, *24*, 36–376. [CrossRef] [PubMed]
13. Kume, H.; Hino, K.; Galipon, J.; Ui-Tei, K. A-to-I editing in the miRNA seed region regulates target mRNA selection and silencing efficiency. *Nucleic Acids Res.* **2014**, *42*, 10050–10060. [CrossRef]
14. Zhang, L.; Yang, C.S.; Varelas, X.; Monti, S. Altered RNA editing in 3’ UTR perturbs microRNA-mediated regulation of oncogenes and tumor-suppressors. *Sci. Rep.* **2016**, *6*, 23226. [CrossRef]
15. Prasanth, K.V.; Prasanth, S.G.; Xuan, Z.; Hearn, S.; Freier, S.M.; Bennett, C.E.; Zhang, M.Q.; Spector, D.L. Regulating gene expression through RNA nuclear retention. *Cell* **2005**, *123*, 249–263. [CrossRef] [PubMed]
16. Goldberg, L.; Abutbul-Amitai, M.; Paret, G.; Nevo-Caspi, Y. Alternative Splicing of STAT3 Is A Target of miR-130a Regulating Lipid Intermediates. *Int. J. Mol. Sci.* **2018**, *19*, 501. [CrossRef] [PubMed]
17. Kitessa, S.M.; Abeywardena, M.Y. Lipid-Induced Insulin Resistance in Skeletal Muscle: The Chase for the Culprit Goes from Total Intramuscular Fat to Lipid Intermediates, and Finally to Species of Lipid Intermediates. *Nutrients* **2016**, *8*, 246. [CrossRef]
18. Li, Q.; Huang, Z.; Zhao, W.; Li, M.; Li, C. Transcriptome Analysis Reveals Long Intergenic Non-Coding RNAs Contributed to Intramuscular Fat Content Differences between Yorkshire and Wei Pigs. *Int. J. Mol. Sci.* **2020**, *21*, 1732. [CrossRef]
19. Liu, K.; Yu, W.; Wei, W.; Zhang, X.; Tian, Y.; Sherif, M.; Liu, X.; Dong, C.; Wu, W.; Zhang, L. Melatonin reduces intramuscular fat deposition by promoting lipolysis and increasing mitochondrial function. *J. Lipid Res.* **2019**, *60*, 767–782. [CrossRef] [PubMed]
20. Zhang, H.; Chen, X.; Sairam, M.R. Novel genes of visceral adiposity: Identification of mouse and human mesenteric estrogen-dependent adipose (MEDA)-4 gene and its adipogenic function. *Endocrinology* **2012**, *153*, 2665–2676. [CrossRef] [PubMed]
21. Liu, X.; Wei, S.; Deng, S.; Li, D.; Liu, K.; Shan, B.; Shao, Y.; Wei, W.; Chen, J.; Zhang, L. Genome-wide identification and comparison of mRNAs, lncRNAs and circRNAs in porcine intramuscular, subcutaneous, retroperitoneal and mesenteric adipose tissues. *Anim. Genet.* **2019**, *50*, 228–241. [CrossRef] [PubMed]
22. Du, J.; Xu, Y.; Zhang, P.; Zhao, X.; Gan, M.; Li, Q.; Ma, J.; Tang, G.; Jiang, Y.; Wang, J. MicroRNA-125a-5p Affects Adipocytes Proliferation, Differentiation and Fatty Acid Composition of Porcine Intramuscular Fat. *Int. J. Mol. Sci.* **2018**, *19*, 501. [CrossRef] [PubMed]
23. Wei, W.; Sun, W.; Han, H.; Chu, W.; Zhang, L.; Chen, J. miR-130a regulates differential lipid accumulation between intramuscular and subcutaneous adipose tissues of pigs via suppressing PPARG expression. *Gene* **2017**, *636*, 23–29. [CrossRef] [PubMed]
24. Wang, Z.; Lian, J.; Li, Q.; Zhang, P.; Zhou, Y.; Zhan, X.; Zhang, G. RES-Scanner: A software package for genome-wide identification of RNA-editing sites. *GigaScience* **2016**, *5*, 37. [CrossRef] [PubMed]
25. Picardi, E.; Pesole, G. REDItools: High-throughput RNA editing detection made easy. *Bioinformatics* **2013**, *29*, 1813–1814. [CrossRef]
26. Wang, K.; Li, M.; Hakonarson, H. ANNOVAR: Functional annotation of genetic variants from high-throughput sequencing data. *Nucleic Acids Res.* **2010**, *38*, e164. [CrossRef]
29. Mudunuri, U.; Che, A.; Yi, M.; Stephens, R.M. bioDBnet: The biological database network. *Bioinformatics* 2009, 25, 555–556. [CrossRef]

30. Zhou, Y.; Zhou, B.; Pache, L.; Chang, M.; Khodabakhshi, A.H.; Tanaseichuk, O.; Benner, C.; Chanda, S.K. Metascape provides a biologist-oriented resource for the analysis of systems-level datasets. *Nat. Commun.* 2019, 10, 1523. [CrossRef]

31. Stark, C.; Breitkreutz, B.J.; Reguly, T.; Boucher, L.; Breitkreutz, A.; Tyers, M. BioGRID: A general repository for interaction datasets. *Nucleic Acids Res.* 2006, 34, D535–D539. [CrossRef] [PubMed]

32. Li, T.; Wernersson, R.; Hansen, R.B.; Horn, H.; Mercer, J.; Slodkowicz, G.; Workman, C.T.; Rigina, O.; Rapacki, K.; Staerfeldt, H.H. A scored human protein-protein interaction network to catalyze genomic interpretation. *Nat. Methods* 2017, 14, 61–64. [CrossRef]

33. Türei, D.; Korcsmáros, T.; Saez-Rodriguez, J. OmniPath: Guidelines and gateway for literature-curated signaling pathway resources. *Nat. Methods* 2016, 13, 966–967. [CrossRef] [PubMed]

34. Bader, G.D.; Hogue, C.W. An automated method for finding molecular complexes in large protein interaction networks. *BMC Bioinform.* 2003, 4. [CrossRef] [PubMed]

35. Peng, L.; Lee, L.J.; Xiong, H.; Su, H.; Rao, J.; Xiao, D.; He, J.; Wu, K.; Liu, D. Characterization of RNA editome in primary and metastatic lung adenocarcinomas. *Oncotarget* 2017, 8, 11517–11529. [CrossRef] [PubMed]

36. Miranda, K.C.; Huynh, T.; Tay, Y.; Ang, Y.-S.; Tam, Y.-L.; Thomson, A.M.; Lim, B.; Rigoutsos, I. A pattern-based method for the identification of MicroRNA binding sites and their corresponding heteroduplexes. *Cell* 2006, 126, 1203–1217. [CrossRef]

37. Ng, P.C.; Henikoff, S. SIFT: Predicting amino acid changes that affect protein function. *Nucleic Acids Res.* 2003, 31, 3812–3814. [CrossRef]

38. Zeng, Z.; Bromberg, Y. Predicting Functional Effects of Synonymous Variants: A Systematic Review and Perspectives. *Front. Genet.* 2019, 10. [CrossRef]

39. Zuker, M. Mfold web server for nucleic acid folding and hybridization prediction. *Nucleic Acids Res.* 2003, 31, 3406–3415. [CrossRef]

40. Ornitz, D.M.; Itoh, N. The Fibroblast Growth Factor signaling pathway. *Wiley Interdiscip. Rev. Dev. Biol.* 2015, 4, 215–266. [CrossRef]

41. Abudesimu, A.; Adi, D.; Siti, D.; Ma, X.; Liu, F.; Xie, X.; Yang, Y.; Li, X.; Chen, B.; Ma, Y. Association of lipid metabolism relevant gene FBXW7 polymorphism with coronary artery disease in Uygur Chinese population in Xinjiang, China: A case-control. *Int. J. Clin. Exp. Pathol.* 2017, 10, 11179–11187. [PubMed]

42. Silva-Vignato, B.; Coutinho, L.L.; Poleti, M.D.; Cesar, A.S.M.; Moncau, C.T.; Regitano, L.C.A.; Balieiro, J.C.C. Gene co-expression networks associated with carcass traits reveal new pathways for muscle and fat deposition in Nelore cattle. *BMC Genomics* 2019, 20, 32. [CrossRef] [PubMed]

43. Gu, L.; Zhu, Y.; Lin, X.; Tan, X.; Lu, B.; Li, Y. Stabilization of FASN by ACAT1-mediated GNPAT acetylation promotes lipid metabolism and hepatocarcinogenesis. *Oncogene* 2020, 39, 2437–2449. [CrossRef] [PubMed]

44. Yu, M.; Gilbert, S.; Li, Y.; Zhang, H.; Qiao, Y.; Lu, Y.; Tang, Y.; Zhen, Q.; Cheng, Y.; Liu, Y. Association of NCOA3 polymorphisms with Dyslipidemia in the Chinese Han population. *Lipids Health Dis.* 2015, 14, 124. [CrossRef]

45. Yi, X.; Wang, Z.; Ren, J.; Zhuang, Z.; Liu, K.; Wang, K.; He, R. Overexpression of chaperonin containing T-complex polypeptide subunit zeta 2 (CCT6b) suppresses the functions of active fibroblasts in a rat model of joint contracture. *J. Orthop. Surg. Res.* 2019, 14, 125. [CrossRef]

46. Mugabo, Y.; Sadeghi, M.; Fang, N.N.; Mayor, T.; Lim, G.E. Elucidation of the 14-3-3ζ interactome reveals critical roles of RNA-splicing factors during adipogenesis. *J. Biol. Chem.* 2018, 293, 6736–6750. [CrossRef]

47. Chertian, P.T.; Al-Khairi, I.; Sriraman, D.; Al-Enezi, A.; Al-Sultan, D.; AlOtaibi, M.; Al-Enezi, S.; Tuomilehto, J.; Al-Mulla, F.; Abubaker, J.A. Increased Circulation and Adipose Tissue Levels of DNAJC27 and Type 2-Diabetes. *Front. Endocrinol. (Lausanne)* 2018, 9, 423. [CrossRef]

48. Taube, M.; Andersson-Assarsson, J.C.; Lindberg, K.; Pereira, M.J.; Gäbel, M.; Svensson, M.K.; Eriksson, J.W.; Svensson, P.-A. Evaluation of reference genes for gene expression studies in human brown adipose tissue. *Adipocyte* 2015, 4, 280–285. [CrossRef]

49. Wang, T.; Feugang, J.M.; Crenshaw, M.A.; Regmi, N.; Blanton, J.R.; Liao, S.F. A Systems Biology Approach Using Transcriptomic Data Reveals Genes and Pathways in Porcine Skeletal Muscle Affected by Dietary Lysine. *Int. J. Mol. Sci.* 2017, 18, 885. [CrossRef]
50. Wu, R.; Yao, Y.; Jiang, Q.; Cai, M.; Liu, Q.; Wang, Y.; Wang, X. Epigallocatechin gallate targets FTO and inhibits adipogenesis in an mRNA m6A-YTHDF2-dependent manner. *Int. J. Obes.* **2018**, *42*, 1378–1388. [CrossRef]

51. Liu, X.; Trakooljul, N.; Hadlich, F.; Mura, E.; Wimmers, K.; Ponsuksili, S. MicroRNA-mRNA regulatory networking fine-tunes the porcine muscle fiber type, muscular mitochondrial respiratory and metabolic enzyme activities. *BMC Genomics* **2016**, *17*, 531. [CrossRef] [PubMed]

52. Mentzel, C.M.J.; Alkan, F.; Keinicke, H.; Jacobsen, M.J.; Gorodkin, J.; Fredholm, M.; Cirera, S. Joint Profiling of miRNAs and mRNAs Reveals miRNA Mediated Gene Regulation in the Göttingen Minipig Obesity Model. *PLoS ONE* **2016**, *11*, e0167285. [CrossRef] [PubMed]

53. Wu, D.; Xi, Q.Y.; Cheng, X.; Dong, T.; Zhu, X.-T.; Shu, G.; Wang, L.-N.; Jiang, Q.-Y.; Zhang, Y.-L. miR-146a-5p inhibits TNF-α-induced adipogenesis via targeting insulin receptor in primary porcine adipocytes. *J. Lipid Res.* **2016**, *57*, 1360–1372. [CrossRef] [PubMed]

54. Yuan, F.; Ma, J.; Xiang, X.; Lan, H.; Xu, Y.; Zhao, J.; Li, Y.; Zhang, W. Improvement of Adipose Macrophage Polarization in High Fat Diet-Induced Obese GHSR Knockout Mice. *BioMed Res. Int.* **2018**, *2018*, 4924325. [CrossRef] [PubMed]

55. Lv, Y.; Liang, T.; Wang, G.; Li, Z. Ghrelin, a gastrointestinal hormone, regulates energy balance and lipid metabolism. *Biosci. Rep.* **2018**, *38*. [CrossRef]

56. Muhammad, H.F.L. Obesity as the Sequel of Childhood Stunting: Ghrelin and GHSR Gene Polymorphism Explained. *Acta Med. Indones.* **2018**, *50*, 159–164.