Genome-Wide Network of lncRNA–mRNA During Ovine Oocyte Development From Germinal Vesicle to Metaphase II in vitro

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Long non-coding RNA (lncRNA) is involved in many biological processes, and it has been closely investigated. However, research into the role of lncRNA in ovine ovarian development is scant and poorly understood, particularly in relation to the molecular mechanisms of ovine oocyte maturation. In the current study, RNA sequencing was performed with germinal vesicle (GV) and in vitro matured metaphase II (MII) stage oocytes, isolated from ewes. Through the use of bioinformatic analysis, abundant candidate lncRNAs in stage-specific ovine oocytes were identified, and their trans- and cis-regulatory effects were deeply dissected using computational prediction software. Functional enrichment analysis of these lncRNAs revealed that they were involved in the regulation of many key signaling pathways during ovine oocyte development, which was reflected by their targeted genes. From this study, multiple lncRNA–mRNA networks were presumed to be involved in key signaling pathways regarding ovine oocyte maturation and meiotic resumption. In particular, one novel lncRNA (MSTRG.17927) appeared to mediate the regulation of phosphatidylinositol 3-kinase signaling (PI3K) signaling during ovine oocyte maturation. Therefore, this research offers novel insights into the molecular mechanisms underlying ovine oocyte meiotic maturation regulated by lncRNA–mRNA networks from a genome-wide perspective.

**Keywords:** lncRNA–mRNA network, ewe, oocytes, meiotic maturation, RNA-Seq

**INTRODUCTION**

In mammals, fully grown follicles release mature oocytes that are ready for fertilization. Each oocyte is enclosed in a follicle and undergoes an extended period of meiotic arrest that may last 40 years or more in humans and some other animals (Woodruff et al., 2010; Conti et al., 2012), during which time the oocytes remain at the first meiotic prophase. Only when luteinizing
hormones (LH) triggers ovulation in the adolescent do the germinal vesicle (GV) oocytes resume meiosis, complete the first meiosis, extrude a polar body, and then reach the metaphase of the second meiosis (MII) ready for fertilization.

A long period of meiotic prophase or GV arrest is suggested to be of great importance for the production of a fertilization-competent oocyte (Conti et al., 2012; Coticchio et al., 2013). The maintenance of this arrest involves a regulatory network between oocytes and somatic cells, thought to be dependent on high levels of cyclic AMP (cAMP) (Cho et al., 1974; Bornslaeger et al., 1986; Conti et al., 2002, 2012) that can activate protein kinase A (PKA) and contribute to the inactivation of the meiosis-promoting factor (MPF), as cAMP decreases PKA is inactivated leading to the activation of MPF, which is responsible for the resumption of oocyte development and GV break down (GVBD) (Jones, 2004; Coticchio et al., 2013). During arrest, cAMP levels within oocytes usually remain relatively high. cAMP can be specifically degraded by phosphodiesterase 3A (PDE3A); however, PDE3A activity is inhibited by cyclic GMP (cGMP) (Norris et al., 2009; Vaccari et al., 2009) that is produced by cumulus cells and diffuses through gap junctions into the oocyte to prevent initiation of meiotic resumption. Following the LH surge, the secretion of natriuretic peptide precursor type C in mural granulosa cells is prevented, which impairs cGMP synthesis (Zhang M. et al., 2010; Wang et al., 2019); as a result, PDE3A is activated, and cAMP is reduced by phosphodiesterase, which in turn permits the resumption of meiosis.

Non-coding RNAs (ncRNAs) are generally divided into two classes according to length: short/small ncRNAs [such as micro-RNAs (miRNAs)] and long non-coding RNAs (lncRNAs) with a length generally exceeding 200 bp (Nagano and Fraser, 2011; Sun and Kraus, 2013). Following the prevalence of genome-wide transcriptome research, a substantial set of roles for lncRNAs are gradually being exposed. Growing evidence suggests that lncRNAs participate in many biological processes, such as the regulation of gene expression (Kim and Sung, 2012; Rinn and Chang, 2012), cellular differentiation and development (Fatica and Bozzoni, 2014; Liu et al., 2017), apoptosis, and disease (Wapinski and Chang, 2011; Rossi and Antonangeli, 2014; Zhang et al., 2019). Moreover, lncRNAs are known to target chromatin modification complexes that mediate posttranscriptional modifications and translational control of messenger RNA (Fatica and Bozzoni, 2014); they also play a part in genomic reprogramming (Nagano et al., 2008; Loewer et al., 2010). Despite the fact that lncRNAs, such as the well-known H19 and Xist (X inactive-specific transcript), were identified as early as the 1990s (Brockdorff et al., 1991; Carninci et al., 2005), their roles are relatively poorly investigated and understood, particularly in the ovary. Presently, most studies have focused on the association of lncRNAs with ovarian cancer (Norris et al., 2009; Mitra et al., 2017), and only a few reports have investigated their roles in ovarian development and follicle growth. Battaglia et al. (2017) suggest that 41 lncRNAs interact with oocyte miRNAs and are potentially involved in the regulation of folliculogenesis in humans. Meanwhile, Liu et al. (2018) discovered that lncRNAs/transcripts of uncertain coding potential (TU) are highly expressed in goat ovaries in the luteal phase and that this might regulate the synthesis of progesterone.

RNA transcriptome sequencing is now one of the most powerful methods available for the detection and expression analysis of lncRNAs (Mortazavi et al., 2008; Atkinson et al., 2012). In the present study, we explored the lncRNA and coding gene regulatory networks of ovine oocytes from GV to MII stages. Briefly, GV oocytes that were isolated from ewes and MII oocytes matured in vitro were sequenced for the expression of lncRNA. Through RNA data analysis, the stage-specific lncRNAs and coding genes were identified. Furthermore, the co-expression (trans) and co-location (cis) analyses were implemented with differentially expressed lncRNAs (DEls) and genes (DEGs) to explore the regulatory networks involved in oocyte transition from GV to MII stages.

MATERIALS AND METHODS

Ethics Statement

All the experiments in this research followed national guidelines and were in accordance with the Ethical Committee of Anhui Agricultural University. All surgery on ewes had the approval of the Institution of Animal Care and Use Committee, Anhui Agricultural University (no. 2016017), China.

Ovine Oocyte Collection and Maturation in vitro

Ovine oocytes were collected from a slaughterhouse in Ürümqi in Xinjiang, China. They were then transported in saline solution supplemented with penicillin-streptomycin to the laboratory within 4 h. There, follicles of 2–8 mm in diameter were extracted from the washed ovaries. After settlement for a short time, the supernatant was removed, and the cell pellet was resuspended with washing buffer for subsequent dissection and selection under a stereoscopic microscope. The oocytes were divided into two classes: (i) denuded oocytes (GV stage) and (ii) oocytes surrounded by cumulus cells; the latter were treated with hyaluronidase (Sigma, H-3506) to remove the cumulus cells. Oocytes were then cultured in vitro with maturation medium in a humidified incubator at 38.5°C with 5% CO2 for 23–25 h. MII stage oocytes with an extruded first polar body were harvested. The sequence samples were prepared with 30–35 oocytes as one biological replicate.

As in previous reports (Peng et al., 2008; Su et al., 2014), the maturation medium (per 100 ml) for oocytes was composed of double-distilled water, 10 ml fetal bovine serum (FBS; Gibco, 00642), 0.95 g TCM199 (Gibco, 31100-03510), 0.22 g NaHCO3, 0.2384 g HEPES, 0.022 g Na-Py, 0.0075 g penicillin, 0.005 g streptomycin, and 0.003 g glutamine supplemented with 100 IU FSH, 1 ml 17β-E2 (working concentration is 1 μg/ml), and 200 IU LH. Phenol red was added to adjust the pH value, and the medium was filtered through a 0.22 μm filter and preserved at 4°C.
RNA Extraction, Library Construction, and Sample Sequence
RNA samples from the collected oocytes were extracted with TRIzol reagent, and the following pipeline was performed to ensure RNA quantity: 1% agarose gel electrophoresis for RNA degradation and contamination test, NanoPhotometer® spectrophotometry (IMPLEN, Westlake Village, CA, United States) for detecting RNA purity, Qubit® 2.0 Fluorometer (Life Technologies, Camarillo, CA, United States) for quantifying RNA concentration, and RNA integrity assessment with an RNA Nano 6000 Assay Kit and a Bioanalyzer 2100 system (Agilent Technologies, Santa Clara, CA, United States). An aliquot of 3 µg RNA from each sample was used as the input to prepare the sequence. After removal of ribosomal RNA, the sequence libraries were generated with a NEBNext® UltraTM Directional RNA Library Prep Kit (NEB, Ipswich, MA, United States) according to the manufacturer’s recommendations, the quality of which was assessed with an Agilent Bioanalyzer 2100 system. After sample clustering with a TruSeq PE Cluster Kit v3-cBot-HS (Illumina, San Diego, CA, United States) according to the manufacturer’s instructions; sequencing was conducted on an Illumina HiSeq 2000 platform using a 100 bp paired-end strategy.

Pipeline of Data Processing and Analysis
Firstly, quantity control was performed to remove low-quality and adapter reads from the generated raw data using Trimmomatic software (v 0.38) (Bolger et al., 2014). FastQC (v0.11.8) reports were produced to check the obtained data (Andrews, 2010), which would be used for subsequent analyses. With the STAR software (v2.7.1a)1, clean data were aligned to the sheep reference genome (version Oar_v3.1 from ensemble) to produce Bam files. After Bam files were sorted with SAMTools (v1.9) (Li et al., 2009), transcript assembly was performed with StringTie software (v2.0) (Pertea et al., 2015). Subsequently, all samples were merged to generate one GTF file with a setting of a “-m 200” parameter, by which only transcripts with a length >200 bp were retained for the following lncRNA analyses.

Screening for Candidate IncRNAs
We identified candidate IncRNAs to carry out downstream analyses following previously reported procedures (Gao et al., 2017; Sun et al., 2018; Zhang et al., 2019). According to the assembled transcript, preliminary filtration was executed to select non-genetic transcripts with a “classcode” label annotated by GffCompare software2 (v0.11.2). The produced transcript file may then contain the IncRNA transcripts with protein coding potential that should be removed. Here, we predicted the protein coding ability of IncRNA transcripts through three methods: Coding–Non-coding Index (CNCI) (Sun et al., 2013), Coding Potential Calculator (CPC) (Kang et al., 2017), and PfamScan (Finn et al., 2016). The combined analyses guaranteed the accuracy of the final candidate IncRNAs (Figure 1D, marked with black arrow) in our study.

Expression Analysis of mRNA and IncRNA
The expression levels of mRNA and IncRNA were calculated using featureCounts (v1.6.4) (Liao et al., 2013) based on assembled transcript files. Then, the DESeq2 package (Love et al., 2014) was used to normalize expression levels and perform differentially expressed analyses for mRNAs and IncRNAs using R (v 3.6.0). In our research, p-values of the transcripts were adjusted using Benjamini and Hochberg’s approach for controlling the false discovery rate (FDR). Significant DEGs and DELs were determined with padj (adjusted p-value) <0.05.

Co-expression and Co-location Analysis
It is thought that IncRNAs play a trans-role through their related co-expression genes (Yan et al., 2017). In our study, the pairwise significant DEGs and DELs were estimated with Pearson’s correlation coefficient (r) using the R package Hmisc (v4.3)3. Those mRNAs with a p-value < 0.01 and |r-value| > 0.9 were considered as co-expressed genes of their IncRNAs. Further, for cis-regulation, based on the assembled GTF file of candidate IncRNAs, we employed FEELnc (v 0.1.1) (Wucher et al., 2017) to screen the coding genes that were located within 100 kb upstream and downstream of IncRNAs (that is, co-location), and the cis-regulated genes of their IncRNAs were decided with the parameter of “isBest = 1.” According to the illustration of FEELnc software, the IncRNA subtypes were divided into two classes: “Genic” and “Intergenic.” “Genic” included subtypes of “overlapping,” “containing,” and “nested”; “Intergenic” consisted of “same strand,” “convergent,” and “divergent” subtypes. Meanwhile, the core regulatory network was determined with common paired mRNA–IncRNA for co-expression and co-location modules.

Functional Enrichment Analysis
Gene Ontology (GO) and Kyoto Encyclopedia of Genes and Genomes (KEGG) enrichment analyses were performed predominantly based on Metascape4 (Zhou et al., 2019), a web-based portal that operates using gene symbols as inputs and integrates a board of current biological databases. In the output results, it highlights the key and non-redundant functional terms, also the enrichment network displays intra- and inter-cluster connection of terms. Due to the limitations of Metascape, large data sets (>3,000 genes) were processed using the clusterProfiler package (Yu et al., 2012) for enrichment analysis; the top 20 most enriched terms are displayed in order according to “GeneRatio,” and the “p.adjust” value was produced using the “BH” method in the parameter of “pAdjustMethod” (Figure 2). Given the inadequate database of functional analysis in sheep, the ovine gene symbols were converted into murine homologous gene symbol IDs through the g:profile website5 (Raudvere et al., 2019).

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1https://github.com/alexdobin/STAR
2https://github.com/gpertea/gffcompare
3https://github.com/harrelfe/Hmisc
4http://metascape.org/gp/index.html
5https://biit.cs.ut.ee/gprofiler/convert
RESULTS

Sequence and Data Preprocessing of Ovine Oocytes at the GV and MII Stages

In the present study, denuded oocytes (GV stage) and COCs were isolated from ovine ovaries. Cumulus cells were removed from COCs and the isolated oocytes developed to the mature MII phase during a 23–25 h incubation *in vitro*. These harvested oocytes were gathered in two groups, “GV” and “MII,” for RNA sequencing (Figure 1A). Representative images of GV and MII ovine oocytes are shown in Figure 1B. To determine the molecular mechanism of lncRNA and related target genes, with quality control and high mapping rates (around 90%; Supplementary Figure S1A), the clean data were processed to identify candidate lncRNAs and coding gene transcripts according to standard pipelines (Figure 1C). In particular, candidate lncRNAs were generated after protein coding potential assessment with CNCI, CPC2, and PfamScan (Figure 1D with a black arrow). Ultimately, we obtained transcript results of lncRNAs and mRNAs with density peaks of 695.54 and 933.66 bp in length, respectively (Figure 1E), which were used for downstream analyses.

Divergent Expression Patterns of IncRNA and mRNA

According to data processing, more transcripts of coding genes were generated than lncRNAs (Figure 2A). To understand the temporal expression patterns of lncRNA and mRNA in ovine oocyte samples, principal component analysis (PCA) was conducted (Figure 2B); it showed differing expression patterns of lncRNAs and mRNAs in GV and MII oocytes, which suggested that the mRNAs and lncRNAs displayed...
a temporal expression pattern in ovine oocytes dependent on the stage of development. Furthermore, volcano plots demonstrated there were 2,941 DELs (Supplementary Table S1) and 6,549 DEGs (Supplementary Table S2 and Figure 2C) of MII vs. GV stage. Moreover, the number of upregulated lncRNAs and mRNAs were both slightly more than that of downregulated ones (Figure 2D). In addition, the top 40 DELs (Supplementary Figure S1B) and DEGs (Supplementary Figure S1C) were made into a heatmap. In order to explore the biological processes of DEGs, functional enrichment was performed (Figure 2E). GO results showed that the top enriched terms were DNA repair, ncRNA metabolic processes, and ribonucleoprotein complex biogenesis; others were also closely related to cellular transcriptional states, such as mRNA processing, mitotic cell cycle, and chromatin and histone modification. Moreover, KEGG revealed that mitogen-activated protein kinase (MAPK), mammalian target of rapamycin (mTOR), and the insulin signaling pathways were likely involved in the regulation of oocyte fate during transition from GV to MII (Figure 2F).
Co-expression Analysis of DELs and DEGs in Ovine Oocytes

To precisely identify the regulatory mechanisms of lncRNAs and mRNAs, co-expression analysis was executed based on DEGs and DELs as previously reported (Liu et al., 2017; Zhang et al., 2019). After filtration based on p-value ($p < 0.01$, 16.52%) and Pearson’s correlation coefficient ($|r| > 0.9$), 173.268 pairwise lncRNA–mRNAs were produced (Supplementary Table S3), which uniquely included 357 DEGs and 2.935 DELs (Figure 3A); among these DEGs, more genes were downregulated at the MII stage (Supplementary Figure S2A). Further, GO enrichment analysis was applied using the co-expressed DEGs, and several significant GO terms were mostly connected with reproductive processes including calcium ion transport, stress-activated MAPK cascade, cellular response to cAMP, insulin secretion, and phosphatidylinositol 3-kinase signaling (PI3K; Figure 3B), despite the top terms appearing to be unrelated (Supplementary Figure S2B). In addition, the expression levels of genes enriched in the five reproductive terms are shown in the heatmap in Figure 3C, which were under the control of numerous lncRNAs (Supplementary Table S4).

Cis-Regulation of lncRNAs With Target Genes

lncRNAs also export a potential cis-regulatory capacity to their adjacent coding genes (Wang and Chang, 2011; Yan et al., 2017); therefore, FEELnc software was used to detect coding genes adjacent to candidate lncRNAs. We obtained 14.821 pairs of cis-regulatory lncRNA genes with the best match (Supplementary Table S5), which included 5.908 unique lncRNAs. In particular, 2.591 lncRNAs (key cis-lncRNAs) were common with co-expressed lncRNAs (Figure 4A), and almost half of them were increased (Supplementary Figure S3A). According to statistical analysis of the subtype, most key cis-lncRNAs

![Figure 3](image-url)
occupied the same strand as coding genes (Figure 4B), and results of genome location showed that over one third of key cis-lncRNAs were located either downstream or upstream of the target gene (Figure 4C). The analysis also showed that there were 2,039 cis-genes under the control of these cis-lncRNAs. Functional enrichment suggested they gathered in cell projection organization, developmental growth, and anatomical structure size (Figure 4D and Supplementary Figure S3B). Moreover, the most enriched pathways were insulin secretion, estrogen, and cAMP signaling (Figure 5A), which are all involved in reproductive processes. Among them, the cis-regulatory lncRNA–mRNA networks of estrogen (Supplementary Table S6) and cAMP signaling (Supplementary Table S7) were highlighted (Figure 5B).

Identification of Core lncRNA–mRNA Network

In addition, core networks were further identified, derived from common terms of trans- and cis-regulation. In our results, 23 cis-genes overlapped with co-expressed genes (Figure 6A), among which six paired lncRNA–mRNA were both detected by the two
methods (Figure 6B). Of note, only one lncRNA was upregulated (MSTRG.1318); others were all significantly downregulated during the GV-to-MII transition (Figure 6C). The target genes controlled by these core lncRNAs (Supplementary Table S8) mostly were enriched in epiboly and PI3K signaling, while more genes were associated with monovalent inorganic cation transport and transport of small molecules (Figure 6D). Among them, MSTRG.17927 was a candidate for tracking; there were 69 genes (which had annotated gene symbols) under its control (Figure 6E and Supplementary Table S9), which were significantly gathered in PI3K signaling (Figure 6F and Supplementary Figure S3C); among them, 12 genes increased (Figure 6G), and 57 were downregulated (Supplementary Figure S3D) during the GV-to-MII transition.

DISCUSSION

Long non-coding RNAs are associated with many reproductive processes in animals (Golicz et al., 2018), including sex determination, spermatogenesis, and genomic imprinting (Gabory et al., 2009; Zhang L. et al., 2010; Arun et al., 2012). However, their roles in ovarian development are poorly understood. In the present study, genome-wide candidate lncRNAs in GV and MII stage oocytes were identified; and trans- and cis-regulation of lncRNAs were inferred with their targeted coding genes based on DELs and DEGs. Moreover, lncRNA–mRNA networks were assumed to regulate key signaling pathways regarding oocyte meiotic maturation. It is important to note that the mechanisms involved in the in vitro maturation of oocytes are potentially different from those occurring in vivo. Nevertheless, the current study provides a guide to the mechanisms involved in oocyte maturation that are regulated by lncRNA.

As reported, the potential coding capability is an important index, which can be used to distinguish protein-coding genes from non-coding transcripts (Sun et al., 2013; Zhao et al., 2018). In our study, 17,512 candidate lncRNAs were identified using strict screening conditions, and only around 6,600 lncRNAs were expressed in at least one sample. It was found that the length of lncRNA transcripts was shorter than that of coding genes, consistent with many previous reports (Ran et al., 2016; Liu et al., 2017; Yan et al., 2017). Furthermore, the expression of lncRNA usually exerts tissue- and stage-specific patterns in animals (Liu et al., 2017; Yan et al., 2017). Thus, the abundance profiles identified here were assumed to be candidate oocyte-specific lncRNAs. Among these, we detected 2,941 lncRNAs that were differentially expressed between GV and MII stage (Figure 2C); we suggest
FIGURE 6 | Identification of core IncRNA–mRNA networks for the maturation of ovine oocytes. (A) Venn plot of cis-regulation with co-expressed genes; common regions refer to candidate core genes. (B) Candidate core IncRNA–mRNA network. (C) Expression of core IncRNAs from GV to MII stages. (D) Enrichment analysis of genes targeted by core IncRNAs. (E) Core IncRNA–mRNA network of targeted genes and MSTEG 17927; the yellow symbol marks negative regulatory genes, and the cyan marks positive regulatory genes. (F) Functional enrichment analysis of targeted genes of MSTEG 17927. (G) Expression of negative regulatory genes targeted by MSTEG 17927.
that these lncRNAs were possibly related to the specificity of the developmental stage of the ovine oocytes.

Currently, the potential functions of lncRNAs are investigated through their target genes by using trans- and cis-regulatory methods as previously described (Gao et al., 2017; Zhang et al., 2019). In this study, we demonstrated that the co-expressed genes (trans-regulation) might be involved in the functional regulation of calcium ion transport and cAMP response. It has been well demonstrated that calcium in many species is involved in oocyte meiotic maturation (Tosti, 2006; Miao and Williams, 2012). In addition, cAMP is responsible for oocyte meiotic arrest and resumption (Norris et al., 2009; Cao et al., 2018). In this study, RAPGEF3, also termed EPAC1, was identified as an important targeted gene of lncRNAs (Supplementary Table S4). Interestingly, RAPGEF3 expression distinctly decreased from GV to MII stage, and the gene has been shown to serve as a cAMP-stimulated GDP exchange factor, whose mutation is known to increase the affinity for cAMP by around 2.5-fold (Dao et al., 2006). Moreover, GO enrichment analysis revealed that several co-expressed genes were enriched in MAPK and PI3K signaling, both of which are known to participate in oocyte meiotic resumption and maturation (Schmitt and Nebreda, 2002; Liang et al., 2007). In addition, lncRNA has previously been reported to be involved in cAMP and MAPK signaling pathways (Sulayman et al., 2019; Wang, 2019). Thus, we suggested that many lncRNAs might participate in oocyte maturation via trans-regulation.

For cis-regulation, the coding genes within 100 kb upstream and downstream of lncRNAs were predicted by FEELnc software. Our results indicated that key cis-lncRNAs mostly were enriched at the subtypes of “same strand” and “divergents”. The majority of cis-lncRNAs are “intergenic” lncRNAs in our study; accordingly, it suggests the intergenic lncRNAs are likely related to modified chromatin as seen in previous reports (Khalil et al., 2009; Ransohoff et al., 2018), which implies that the cis-lncRNA is potentially related to meiotic maturation via chromatin modification. The cis-regulatory effects of lncRNAs on neighboring genes have been well examined (Sun and Kraus, 2013; Yan et al., 2017). Here, with the enrichment of cis-regulated genes, we uncovered the cis-regulatory pathway of lncRNAs; more specifically, two lncRNA–mRNA networks were highlighted (estrogen and cAMP signaling). Furthermore, several cis-regulatory pathways of key lncRNAs were identified, which included insulin secretion, Wnt, Hippo, cGMP-PKG, and calcium signaling (Figure 5A). These results indicated that cis-regulatory lncRNAs likely account for a broad range of biological processes in oocytes during the GV-to-MII stage transition. Importantly, the core lncRNA–mRNA networks were identified by overlapping of the trans- and cis-methods. It is a remarkable point that the lncRNA MSTRG.17927, which declined significantly in MII stage oocytes, is thought to be involved in oocyte maturation via PI3K signaling. Matten et al. (1996) reported that the activation of a MAPK cascade is related to MPF activation, and PI3K has been suggested to be essential for insulin-induced oocyte meiotic resumption (Bagowski et al., 2001; Liang et al., 2007). Interestingly, the co-expressed genes in this study were mostly enriched in insulin secretion. It is also known that preovulatory follicles have an abundance of insulin compared with subordinate follicles in cattle (Landau et al., 2000). Insulin exposure affects meiotic chromatin during oocyte maturation and meiotic progression in mice (Acevedo et al., 2007; Das and Arur, 2017), indicating that insulin plays a role in oocyte maturation. However, lncRNA-mediated insulin secretion in oocyte maturation needs further investigation.

In summary, RNA sequencing was performed with isolated GV and in vitro matured MII stage oocytes. Through prediction analysis, genome-wide candidate lncRNAs that were stage-specific were identified, and their trans- and cis-regulatory functions were investigated. As a result, we suggested that multiple lncRNA–mRNA networks may be involved in key signaling pathways throughout oocyte differentiation during the GV-to-MII stage transition. In particular, MSTRG.17927 was shown to mediate the regulation of PI3K signaling. This research provides novel bioinformatic insights into the role of lncRNA in the molecular mechanisms underlying oocyte development.

DATA AVAILABILITY STATEMENT

The datasets presented in this study can be found in online repositories. The names of the repository/repositories and accession number(s) can be found below: https://www.ncbi.nlm.nih.gov/, GSE148022.

ETHICS STATEMENT

The animal study was reviewed and approved by the Institution of Animal Care and Use Committee, Anhui Agricultural University, China.

AUTHOR CONTRIBUTIONS

J-JW, M-HN, and TZ conducted the experiments. J-JW analyzed the data. J-JW, H-GC, and WS wrote the manuscript. H-GC and WS designed the experiments. All authors reviewed the manuscript.

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SUPPLEMENTARY MATERIAL

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fphys.2020.01019/full#supplementary-material

FIGURE S1 | Statistical data of RNA sequence and expression of differentially expressed genes (DEGs) and differentially expressed lncRNAs (DELs). (A) Table of quality control and mapping rate statistics of RNA sequences; Heatmap of top 40 DELs (B) and DEGs (C).

FIGURE S2 | Co-expression analysis of DELs and DEGs. (A) Heatmap of unique co-expressed DEGs; (B) GO enrichment analysis of unique co-expressed DEGs.

FIGURE S3 | Cis-regulatory analysis of DELs and DEGs. (A) Heatmap of genes targeted by core cis-lncRNAs; (B) Functional enrichment of genes targeted by core cis-lncRNA. (C) Functional enrichment of targeted genes of MSTEG 17927; (D) Expression of positive regulatory genes targeted by MSTEG 17927.

TABLE S1 | Results of differential expression analysis of lncRNAs.
TABLE S2 | Results of differential expression analysis of mRNAs.
TABLE S3 | Results of co-expression analysis.
TABLE S4 | Co-expressed gene pairs of reproduction related GO terms.
TABLE S5 | Classes of lncRNAs for co-location.
TABLE S6 | Estrogen cis-gene pathway.
TABLE S7 | cAMP cis-gene pathway.
TABLE S8 | Candidate core lncRNA-mRNA networks.
TABLE S9 | Target genes of MSTGR.17927.

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