# **MOLECULAR BIOLOGY**

e-ISSN 1643-3750 © Med Sci Monit, 2015; 21: 3449-3458 DOI: 10.12659/MSM.894638

Received: Accepted: Published:	2015.05.13 2015.07.08 2015.11.10		Potential Role of IncRN/ to Pathogenesis of Inter Degeneration Based on	As in Contributing rvertebral Disc Microarray Data			
Authors' Contribution:       BC         Study Design A       BD         Data Collection B       BC         Statistical Analysis C       BC         Data Interpretation D       C         Manuscript Preparation E       AEF         Literature Search F       F         Funds Collection G       F		BC BD BC C AEF F F	Yu Chen* Haijian Ni* Yingchuan Zhao* Kai Chen Ming Li Cheng Li Xiaodong Zhu Qiang Fu	Department of Orthopaedics, Changhai Hospital, Second Military Medical University, Shanghai, P.R. China			
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Background: Material/Methods:		ground: Nethods:	Our study intended to identify potential long non-coding RNAs (IncRNAs) and genes, and to elucidate the un- derlying mechanisms of intervertebral disc degeneration (IDD). The microarray of GSE56081 was downloaded from the Gene Expression Omnibus database, including 5 hu- man control nucleus pulposus tissues and 5 degenerative nucleus pulposus tissues, which was on the basis of GPL15314 platform. Identification of differentially expressed IncRNAs and mRNAs were performed between the 2 groups. Then, gene ontology (GO) and pathway enrichment analyses were performed to analyze the biological functions and pathways for the differentially expressed mRNAs. Simultaneously, IncRNA-mRNA weighted coex- pression network was constructed using the WGCNA package, followed by GO and KEGG pathway enrichment analyses for the genes in the medules. Finally, the protein interaction (DDI) approximately and the protein interaction (DDI) approximately and the protein interaction (DDI).				
Results:		Results:	A total of 135 significantly up- and 170 down-regulated lncRNAs and 2133 significantly up- and 1098 down-regulated mRNAs were identified. Additionally, <i>UBA52</i> (ubiquitin A-52 residue ribosomal protein fusion product 1), with the highest connectivity degree in PPI network, was remarkably enriched in the pathway of me- tabolism of proteins. Eight lncRNAs – LINC00917, CTD-2246P4.1, CTC-523E23.5, RP4-639J15.1, RP11-363G2.4, AC005082.12, MIR132, and RP11-38F22.1 – were observed in the modules of lncRNA-mRNA weighted coex- pression network. Moreover, <i>SPHK1</i> in the green-yellow module was significantly enriched in positive regula- tion of cell migration.				
Conclusions:		lusions:	LncRNAs LINC00917, CTD-2246P4.1, CTC-523E23.5, RP4-639J15.1, RP11-363G2.4, AC005082.12, MIR132, and RP11-38F22.1 were differentially expressed and might play important roles in the development of IDD. Key genes, such as <i>UBA52</i> and <i>SPHK1</i> , may be pivotal biomarkers for IDD.				
MeSH Keywords:		ywords:	Afferent Pathways • Gene Ontology • Intervertebral Disc Degeneration • RNA, Long Noncoding				
	Full-t	ext PDF:	http://www.medscimonit.com/abstract/index/idArt/894638				
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### Background

Intervertebral disc degeneration (IDD) is an important cause of low back pain, which affects patient quality of life and causes heavy socioeconomic burdens [1,2]. Currently, low back pain is mainly treated by medication combined with physiotherapy. Despite progress in surgical therapy, it is not curative and is related with several complications [3]. Fortunately, gene therapy has attracted more attention in recent years [4].

Several studies have unveiled a large number of genes or proteins that are implicated in IDD. For example, a former study has demonstrated that axin2 and Caveolin-1 are down-regulated in IDD [5]. Another study also suggests that expression of IL-1 $\beta$  is significantly higher in IDD [6], and IL-1 $\beta$  provokes the expression of brain-derived neurotrophic factor (BDNF) and vascular endothelial growth factor (VEGF), bringing about angiogenesis in human IDD [7]. Moreover, growth factors including BMP2 as well as TGF- $\beta$ 1 have recently been suggested to enhance the generation of proteoglycans by nucleus pulposus (NP) cells [8]. Degradation of metalloproteinases (MMPs) has been indicated to be connected with IDD [9]. Using a combination of a variety of growth factors has also been proposed to be appropriate to treat IDD, and platelet-rich plasma (PRP) is the source of these growth factors and has inhibitory effects on IDD in a rabbit model [10,11]. Targeting dysregulated genes has been suggested in animal IDD models, with inspiring results [12]. Nevertheless, dysregulation of gene expression is very complicated. Therefore, it is of great importance to elucidate the mechanisms underlying IDD at molecular levels.

Recently, aberrant expression of long non-coding RNAs (InCRNAs) is shown to cause disordered gene expression. LnCRNAs are mRNAlike transcripts ranging in length from 200 nt to 100 kb, and they lack significant open reading frames [13,14]. In spite of this, accumulating studies have indicated that InCRNAs play crucial roles in many biological processes and human diseases [15,16]. For example, Wan et al. [17] demonstrated that over-expressed InCRNA RP11-296A18.3 induced the up-regulation of *FAF1*, which ultimately enhanced the abnormal apoptosis of disc cells. Wan et al. [17] also suggested the crucial roles of differentially expressed InCRNA in the nucleus pulposus of IDD, in which InCRNA-mRNA microarray profile GSE56081 was used. However, there are few studies investigating the role of InCRNAs in IDD. As a result, identifying the aberrant InCRNAs in IDD is still quite a challenge.

In our study, we intended to use the same microarray profile to further identify the differentially expressed lncRNA and mRNAs. Moreover, gene ontology (GO), Kyoto Encyclopedia of Genes and Genomes (KEGG) pathway, construction of lncRNA-mRNA weighted coexpression network, and protein-protein interaction (PPI) network were used to elucidate the molecular mechanism in the process of IDD. We believe that different results obtained from the microarray profile of Wan et al. might provide a profound understanding of IDD.

### **Material and Methods**

#### Tissue samples and data acquisition

The gene expression data of GSE56081 deposited by Wan et al. [17] was downloaded from the National Center of Biotechnology Information (NCBI) Gene Expression Omnibus (*http://www.ncbi.nlm.nih.gov/geo/*) database, which was on the basis of GPL15314 platform of Arraystar Human LncRNA microarray V2.0 (Agilent\_033010 Probe Name version). A total of 10 samples (5 human control nucleus pulposus tissues and 5 degenerative nucleus pulposus tissues) were included for the development of this microarray.

#### **Data preprocessing**

Expression data of probes were converted to corresponding genes symbols according to the annotation of GPL15314 platform. Aggregate() function of R was used to compute the average expression value, which was used for genes corresponding to multi-probes, and KNN method [18] of Impute package of R [19] was used to add the missing values of probes.

Then, quantile normalization was carried out using preprocessCore package of R [20]. Subsequently, the expression matrix was obtained.

# Identification of differentially expressed lncRNAs and mRNAs

Limma (linear models for microarray data) package [21] in R was used to identify differentially expressed lncRNAs and mRNAs between the 2 groups through use of the *t* test. Multiple test-ing correction was implement by calculating the Benjamini-Hochberg [22] false discovery rate (FDR). |log2 fold-change| >1 and an FDR <0.05 were regarded as the criteria for differential expression.

#### Gene ontology (GO) and pathway enrichment analysis

GO analysis is frequently used in functional enrichment studies of large-scale genes [23]. Kyoto Encyclopedia of Genes and Genomes (KEGG) enrichment analysis was performed to analyze the biological pathways, involving the differentially expressed mRNAs. In the current study, TargetMine (*http://targetmine.nibio.go.jp*) [24] was used to investigate the functional enrichment condition for the up- and down-regulated differentially expressed mRNAs. FDR less than 0.05 was selected as the threshold.



Figure 1. (A) Box plot of gene expressions in human control nucleus pulposus tissue and degenerative nucleus pulposus tissue before normalization. (B) Box plot of gene expressions in human control nucleus pulposus tissue and degenerative nucleus pulposus tissue after normalization. The X axis represents samples and the Y axis represents expression level of genes after log2transformation. The black line in the center is the median of expression value, and consistent distribution implies a good normalization.

# The construction of lncRNA-mRNA weighted coexpression network

WGCNA package [25] in R was used to construct the lncRNA/ mRNA coexpression network. The network construction steps included the following: (1) Network construction: the lncRNAmRNA weighted coexpression network is fully specified by its adjacency matrix  $a_{mn}$ , and  $a_{mn}$  encodes the network connection strength between nodes m and n. In order to calculate the adjacency matrix, the default approach defines the co-expression similarity S<sub>mn</sub> as the absolute value of the correlation coefficient between nodes of m and n:  $S_{mn} = |cor(a_m, a_n)|$ . The weighted adjacency am between 2 genes is proportional to their similarity on a logarithmic scale with  $\beta \ge 0.8$ ,  $\log(a_{mn}) = \beta \times \log(S_{mn})$ . Adjacency functions were obtained by approximate scalefree topology criterion. Then, the adjacency matrix was converted into topology matrix; (2) Module detection: we applied Dynamic Tree Cut method and Static Tree Cut method to detect modules with a minimum of 30 lncRNA/genes; and (3) Correlation analysis between disease information and gene module using the correlation coefficient method and gene significance (GS) value.

Then, we performed the GO and KEGG pathway enrichment analysis of genes in the module.

#### **PPI network construction**

The interaction of the proteins were analyzed by means of the online STRING database (Search Tool for the Retrieval of





Interacting Genes, *http://string-db.org/*) [26] and the combined score >0.9 was used as the cut-off criterion. Subsequently, Cytoscape (*http://cytoscapeweb.cytoscape.org/*) [27] was used to display the PPI network.

Ontology	Term	FDR	Count
GO terms			
GO-BP	Metabolic process	2.55×10⁻²	827
GO-BP	Macromolecule metabolic process	1.41×10 <sup>-3</sup>	650
GO-BP	Cellular macromolecule metabolic process	1.28×10⁻³	610
GO-BP	Cellular component organization or biogenesis	5.10×10 <sup>-3</sup>	471
GO-BP	Cellular component organization	1.33×10⁻³	468
GO-CC	Intracellular	2.64×10 <sup>-7</sup>	972
GO-CC	Intracellular part	9.77×10⁻ <sup>7</sup>	963
GO-CC	Organelle	1.74×10 <sup>-7</sup>	946
GO-CC	Membrane-bounded organelle	1.13×10 <sup>-8</sup>	903
GO-CC	Intracellular organelle	2.34×10 <sup>-7</sup>	837
GO-MF	Binding	1.03×10 <sup>-2</sup>	904
GO-MF	Organic cyclic compound binding	3.76×10 <sup>-2</sup>	309
GO-MF	Heterocyclic compound binding	2.22×10 <sup>-2</sup>	305
GO-MF	Nucleic acid binding	9.35×10⁻⁴	262
GO-MF	RNA binding	1.31×10 <sup>-5</sup>	185
KEGG pathways			
	Gene expression	2.44×10⁻⁵	168
	Metabolism of proteins	9.99×10⁻ <sup>8</sup>	116
KEGG	Cellular responses to stress	6.40×10 <sup>-10</sup>	64
	Translation	2.24×10 <sup>-15</sup>	52
	Alcoholism	8.50×10 <sup>-10</sup>	50

 
 Table 1. The top 5 Gene Ontology (GO) functions and Kyoto Encyclopedia of Genes and Genomes (KEGG) pathways enrichment analysis of up-regulated differentially expressed genes.

BP – biological process; CC – cellular components; MF – molecular function.

Table 2. The results of correlation analysis based on correlation coefficient method.

Module	ME purple	ME greenyellow	ME magenta	ME yellow	ME green	ME grey60	ME grey
Significant	-0.81	-0.85	-0.85	-0.86	-0.9	-0.97	-0.034
P value	4.93×10 <sup>-3</sup>	1.80×10 <sup>-3</sup>	1.67×10 <sup>-3</sup>	1.29×10 <sup>-3</sup>	4.65×10 <sup>-4</sup>	4.19×10 <sup>-6</sup>	9.27×10 <sup>-1</sup>

### Results

#### Data preprocessing

On the basis of the annotation information of GPL56081, a total of 18 208 probes were identified, including 1970 lncRNAs and 16238 mRNAs. The data before and after normalization are shown in Figure 1A and 1B.

# Identification of differentially expressed lncRNAs and mRNAs

Based on the cut-off criteria, 135 significantly up- and 170 down-regulated lncRNAs were screened out. Moreover, 2133

significantly up- and 1098 down-regulated mRNAs were observed, as shown in Figure 2.

#### GO and pathway enrichment analysis

TargetMine was used to identify GO enriched functions for significant differentially expressed mRNAs. Top 5 GO and KEGG terms are listed in Table 1. Up-regulated genes were significantly enriched in macromolecule metabolic process (FDR=1.41E-03) and cellular macromolecule metabolic process (FDR=1.28E-03) of BP, membrane-bounded organelle (FDR=1.13E-08) and organelle (FDR=1.74E-07) of CC, RNA binding (FDR=1.31E-05), and nucleic acid binding (FDR=9.35E-04).

Significantly enriched KEGG pathways of up-regulated differentially expressed mRNAs were mainly on translation (FDR=2.24E-15) and cellular responses to stress (FDR=6.40E-10). However, the significant functions and pathways of down-regulated genes were not identified.

# The construction of lncRNA-mRNA weighted coexpression network

In the current study, a total of 7 weighted coexpression subnetworks were identified. The results of correlation analysis demonstrated that in addition to the grey module, the other 6 modules were highly relevant in intervertebral disc degeneration, as shown in Table 2 and Figure 3. Then, IncRNA-mRNA weighted coexpression network was constructed on the basis of the genes with top 30 connectivity degrees in 6 modules and P value <0.01. Among these, 8 lncRNAs were involved in the 4 modules (green module: CTC-523E23.5, RP4-639J15.1 and RP11-363G2.4; green-yellow module: CTD-2246P4.1 and LINC00917; magenta module: AC005082.12; purple module: MIR132 and RP11-38F22.1, Table 3) and were up-regulated (Figure 4A-4D). Notably, in these modules, red diamond nodes represented up-regulated lncRNA, pink square nodes represented up-regulated genes, and pastel green square nodes represented down-regulated genes (Figure 4).

Then, we performed the GO and KEGG pathway enrichment analysis of genes in the module. As shown in Table 4, the genes in the green module were significantly enriched in regulation of cell adhesion and response to axon injury. The genes in the green-yellow module were significantly enriched in positive regulation of cell migration and positive regulation of cell motion. The genes in the magenta module were significantly enriched in nucleotide catabolic process as well as nucleobase, nucleoside, nucleotide, and nucleic acid catabolic process. The genes

#### Table 3. The information of key lncRNAs in different modules.



Figure 3. Significant modules in the IncRNA-mRNA weighted coexpression network based on the gene significance (GS) value.

in the purple module were significantly enriched in contractile fiber part and striated muscle thin filament.

#### **PPI network**

The PPI network of significantly up-regulated genes was constructed using the STRING database. As shown in Figure 5, several PPI nodes were high in connectivity degrees, as follows: *UBA52* (ubiquitin A-52 residue ribosomal protein fusion product 1, degree=51), RPS3 (ribosomal protein S3, degree=41), *RPL9* (ribosomal protein L9, degree=38), *RPL19* (ribosomal protein L19, degree=37), *RPL23* (ribosomal protein L23, degree=36), *RPL30* (ribosomal protein L30, degree=36), and *RPS14* (ribosomal protein S14, degree=36).

Module	Gene id	Gene name	Chromesome	log2FC
Purple	ENSG00000267200.1	MIR132	chr17	1.47
Green	ENSG00000269086.2	CTC-523E23.5	chr19	1.81
Green	ENSG00000236102.2	RP4-639J15.1	chr7	1.96
Green yellow	ENSG00000267065.1	CTD-2246P4.1	chr17	1.53
Green	ENSG00000229556.2	RP11-363G2.4	chr13	1.90
Magenta	ENSG00000226816.2	AC005082.12	chr7	2.13
Green yellow	ENSG00000168367.5	LINC00917	chr16	1.41
Purple	ENSG00000258220.1	RP11-38F22.1	chr12	1.90

FC - fold change.



Figure 4. The sub-network of genes with top 30 connectivity degrees in green module (A), green-yellow module (B), magenta module (C), and purple module (D). Red diamond nodes stand for up-regulated lncRNA; pink nodes stand for up-regulated genes; pastel green nodes stand for down-regulated genes.

#### Discussion

In this study, we identified 135 significantly up- and 170 downregulated lncRNAs and 2133 significantly up- and 1098 down-regulated mRNAs. Additionally, up-regulated *UBA52* had the highest connectivity degree in PPI network and was remarkably enriched in the pathway of metabolism of proteins. IncRNA-mRNA weighted coexpression network analysis showed that a total of 8 key lncRNAs were identified and were remarkably higher in IDD samples relative to normal tissues, for example, lncRNA LINC00917 and CTD-2246P4.1 were identified in the green-yellow module. These lncRNAs and their interacting genes may be responsible for IDD development and merit further consideration.

Significantly, up-regulated *UBA52* had the highest connectivity degrees in PPI network and was remarkably enriched in the pathway of metabolism of proteins in our study. The intervertebral disc (IVD), a specialized fibrocartilaginous connective tissue, is mainly composed of the cell and extracellular matrix (ECM), and the molecular components of ECM are collagen, elastin, and proteoglycans [28]. Generally, the destruction of ECM is believed to be an important cause of IDD because the degradation of ECM help endothelial cells to migrate to further induce the generation of neovascularization [29,30]. MMPs are demonstrated to be the major proteolytic enzymes responsible for destruction of the ECM in the IVD [31]. UBA52 is a fusion protein composed of ubiquitin and ribosomal protein. Moreover, ubiquitin proteins are encoded by some genes, including UBA52 and UBA80, and are typically related with tagging proteins for proteosome degradation [32,33]. A previous study has demonstrated that extracellular ubiquitin increases MMP-2 protein level and enhances the migration of

Table 4. The GO and KEGG pathways enrichment analysis of differentially expressed genes.

Category	Term	Count	P value	Genes	
Green module					
GO_BP	Regulation of cell adhesion	3	3.03×10 <sup>-2</sup>	LAMA2, S1PR1, JAK2	
GO_BP	Response to axon injury	2	3.92×10 <sup>-2</sup>	TXN2, JAK2	
GO_BP	Regulation of cell migration	3	4.44×10 <sup>-2</sup>	LAMA2, S1PR1, JAK2	
GO_BP	Activation of MAPKK activity	2	4.88×10 <sup>-2</sup>	ZAK, JAK2	
GO_CC	Integral to plasma membrane	6	4.76×10 <sup>-2</sup>	STX8, TSPAN31, ICAM3, ITGA11, HLA-DPA1, KCNJ2	
REACTOME_ PATHWAY	Integrin cell surface interactions	3	1.43×10 <sup>-2</sup>	LAMA2, ICAM3, ITGA11	
Greenyellow r	nodule				
GO_BP	Positive regulation of cell migration	3	8.29×10⁻³	EGFR, SPHK1, F7	
GO_BP	Positive regulation of cell motion	3	9.97×10⁻³	EGFR, SPHK1, F7	
GO_BP	Positive regulation of locomotion	3	9.97×10⁻³	EGFR, SPHK1, F7	
GO_BP	Regulation of ATPase activity	2	2.46×10 <sup>-2</sup>	TPM2, TNNI3	
GO_BP	Positive regulation of cell proliferation	4	2.51×10 <sup>-2</sup>	EGFR, FGF7, SPHK1, SHC1	
GO_MF	Cytoskeletal protein binding	5	7.94×10 <sup>-3</sup>	EGFR, SUN2, FXYD5, TPM2, TNNI3	
GO_MF	Actin binding	4	1.49×10 <sup>-2</sup>	EGFR, FXYD5, TPM2, TNNI3	
GO_CC	Shc-EGFR complex	2	2.97×10⁻³	EGFR, SHC1	
GO_CC	Striated muscle thin filament	2	2.21×10 <sup>-2</sup>	TPM2, TNNI3	
Magenta mod	lule				
GO_BP	Nucleotide catabolic process	3	3.95×10⁻³	SMUG1, NTHL1, AMPD3	
GO_BP	Nucleobase, nucleoside, nucleotide and nucleic acid catabolic process	3	5.09×10 <sup>-3</sup>	SMUG1, NTHL1, AMPD3	
GO_BP	Nucleobase, nucleoside and nucleotide catabolic process	3	5.09×10 <sup>-3</sup>	SMUG1, NTHL1, AMPD3	
GO_BP	Nitrogen compound catabolic process	3	6.74×10 <sup>-3</sup>	SMUG1, NTHL1, AMPD3	
GO_BP	Heterocycle catabolic process	3	8.81×10 <sup>-3</sup>	SMUG1, NTHL1, AMPD3	
GO_MF	Oxidized base lesion DNA N-glycosylase activity	2	5.77×10 <sup>-3</sup>	SMUG1, NTHL1	
GO_MF	DNA N-glycosylase activity	2	2.10×10 <sup>-2</sup>	SMUG1, NTHL1	
GO_MF	Hydrolase activity, hydrolyzing N-glycosyl compounds	2	3.78×10 <sup>-2</sup>	SMUG1, NTHL1	
Purple module					
GO_CC	Contractile fiber part	9	4.26×10 <sup>-3</sup>	ACTG2, TNNT3, ACTA1, MYL2, TNNC2, MYL3, TNNC1, ANKRD1, LMAN1	
GO_CC	Striated muscle thin filament	4	4.34×10 <sup>-3</sup>	TNNT3, ACTA1, TNNC2, TNNC1	
GO_CC	Contractile fiber	9	6.42×10 <sup>-3</sup>	ACTG2, TNNT3, ACTA1, MYL2, TNNC2, MYL3, TNNC1, ANKRD1, LMAN1	
GO_CC	Sarcomere	8	7.00×10 <sup>-3</sup>	TNNT3, ACTA1, MYL2, TNNC2, MYL3, TNNC1, ANKRD1, LMAN1	

Category	Term	Count	P value	Genes
GO_CC	Actin cytoskeleton	14	8.75×10 <sup>-3</sup>	PRKCZ, MYL2, HAX1, ACTA1, TNNC2, MYO1B, TNNC1, MYL3, SNCA, VILL, AMPH, LSP1, ACTG2, TNNT3
GO_BP	Regulation of megakaryocyte differentiation	3	5.55×10 <sup>-3</sup>	TESC, HIST1H4L, PF4, HIST1H4D
GO_BP	Negative regulation of protein metabolic process	12	5.81×10 <sup>-3</sup>	BAK1, PRKCZ, SET, EIF4EBP2, ANAPC5, SERPINA5, SNCA, CDC23, ANKHD1, MDM4, EIF2B3, PSMB9
GO_BP	Negative regulation of cellular protein metabolic process	11	1.22×10 <sup>-2</sup>	BAK1, PRKCZ, SET, EIF4EBP2, ANAPC5, SERPINA5, SNCA, CDC23, ANKHD1, EIF2B3, PSMB9
GO_BP	Ventricular cardiac muscle morphogenesis	4	1.53×10 <sup>-2</sup>	MYL2, MYL3, TNNC1, COL11A1
GO_BP	Nucleosome assembly	7	1.60×10 <sup>-2</sup>	HIST1H2AB, HIST1H2BB, HIST1H4L, TSPYL2, SET, HIST1H1A, HIST1H3E, HIST1H4D, HIST1H3G
GO_MF	Carbohydrate binding	16	2.02×10 <sup>-2</sup>	EPM2A, HEXB, HK2, CHI3L1, POSTN, PF4, LMAN2, GYG1, LMAN1, ITGAM, TNFAIP6, SERPINA5, VEGFA, PKD1, FGF1, SELPLG
KEGG_ PATHWAY	Cardiac muscle contraction	8	3.55×10⁻³	UQCRC1, SLC9A6, ATP2A2, MYL2, MYL3, TNNC1, CACNG6, COX6C
KEGG_ PATHWAY	Systemic lupus erythematosus	7	4.03×10 <sup>-2</sup>	HIST1H2AB, HIST1H2BB, TROVE2, HIST1H4L, HLA-DRB4, HIST1H3E, HIST1H4D, HIST1H3G, CTSG

Table 4 continued. The GO and KEGG pathways enrichment analysis of differentially expressed genes.

endothelial cells [34]. In light of these results, we infer that inhibiting the expression of *UBA52* may be a good strategy in the treatment of IDD by reducing the ubiquitin proteins levels to further suppress the destruction of ECM.

In the green-yellow module, IncRNA LINC00917 and CTD-2246P4.1 were remarkably higher in IDD samples relative to normal tissues. Moreover, their interacted genes, such as upregulated SPHK1 and down-regulated EGFR, were significantly enriched in positive regulation of cell migration. Several growth factors have been demonstrated to increase endothelial cell migration to further induce the generation of neovascularization [30]. Significantly, the loss of ECM and infiltration of blood vessels are 2 characteristics of IDD [28]. SPHK1 is a members of the SPHKs family, which has been indicated to play a crucial role in cell migration [35]. Moreover, growing evidence has demonstrated that SPHK1 participates in the production of angiogenesis. SPHK1 overexpression promoted endothelial cells survival in the absence of extracellular matrix and played a crucial role in vascular generation [36]. In the current study, the level of SPHK1 was up-regulated. Accordingly, SPHK1 might be a potential biomarker in the process of IDD through increasing endothelial cell migration to further induce the generation of neovascularization. Several previous studies have indicated that the abnormal expression of lncRNAs may

dysregulate the expression of their neighboring genes [37]. Our findings suggest that the aberrant expression of *SPHK1* may be caused by the up-regulated LINC00917 or CTD-2246P4.1, to further influence the occurrence of IDD. Therefore, lncRNAs LINC00917 and CTD-2246P4.1 may play a key role in the development of IDD via interacting with *SPHK1*. However, this assumption needs to be further investigated.

Furthermore, our study also identified the other 6 differentially expressed lncRNAs as well as hundreds of differentially expressed genes in different modules. CTC-523E23.5, RP4-639J15.1, and RP11-363G2.4 were identified and could interact with Integrin, Alpha 11 (ITGA11) in the green module. ITGA11 is shown to be differentially expressed and is associated with cell proliferation in IVD [38]. AC005082.12 was screened out and could interact with Ephrin-A3 (EFNA3) in the magenta module. MIR132 and RP11-38F22.1 could interact with Cathepsin L (CTSL) in the purple module. CTSL is demonstrated to degrade ECM and basement membrane components [39]. The destruction of ECM has been regarded to be an important cause of IDD [29,30]. Although the underlying mechanisms of these key IncRNAs in the development of IDD have not been fully discussed in our study, our results indicate that they may contribute to IDD development to some extent.



Figure 5. The protein-protein interaction network of significantly up-regulated genes.

However, there were still some limitations in the current study. First, sample size was small which might limit the validity of the results. Second, no animal experiments were performed in our study to further validate these results. Thus, large sample sizes and more experiments, such as RNA interference and fluorescence *in situ* hybridization, are required in future work to more comprehensively elucidate the mechanisms of lncRNAs in the development of IDD.

### Conclusions

IncRNAs LINC00917, CTD-2246P4.1, CTC-523E23.5, RP4-639J15.1, RP11-363G2.4, AC005082.12, MIR132, and RP11-38F22.1 were differentially expressed and might play important roles in the development of IDD. Key genes, such as *UBA52* and *SPHK1*, may be pivotal biomarkers for IDD. Nevertheless, further experiments are required to validate their effects and mechanisms in IDD.

#### **Conflict of interest**

The authors declare that they have no competing interests.

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