Comprehensive analysis of molecular pathways and key genes involved in lumbar disc herniation

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Abstract

Based on the Thompson classification of intervertebral discs (IVDs), we systematically analyzed gene expression differences between severely degenerated and mildly degenerated IVDs and explored the underlying molecular mechanisms using bioinformatics and multichip integration. We used multomics analysis, including mRNA microarray and methylation chips, to explore the genetic network and mechanisms of lumbar disc herniation (LDH). Subsequently, the Combat function of the R language SVA package was applied to eliminate heterogeneity between the gene expression data. And the protein–protein interaction (PPI) network, gene ontology (GO), and molecular pathways were used to construct the mechanisms network. Consequently, we obtained 149 differentially expressed genes. Related molecular pathways are the following: ribosome activity, oxidative phosphorylation, extracellular matrix response. Besides, through PPI network analysis, genes with higher connectivity such as UBA52, RPLP0, RPL3, RPLP2, and RPL27 were also identified, suggesting that they play important regulatory roles in the complex network associated with LDH. Additionally, cg12556991 (RPL27) and cg06852319 (RPLP0) were found to be LDH-related candidate DNA methylation modification sites in the IVDs tissue of LDH patients. In conclusions, ribosome activity, oxidative phosphorylation, and extracellular matrix response may be potential molecular mechanisms underlying LDH, while hub genes involved in UBA52, RPLP0, RPL3, RPLP2, and RPL27, and candidate DNA methylation modification sites of cg12556991and cg06852319 are likely key regulators in the development of LDH.

Abbreviations: BP = biological processes, CC = cellular components, FC = fold-changes, FDR = false discovery rate, GEO = Gene Expression Omnibus, GO = gene ontology, GWAS = genome-wide association study, IVDs = intervertebral discs, KEGG = Kyoto Encyclopedia of Genes and Genomes, KNN = K-nearest neighbor, LDH = lumbar disc herniation, MALDI = matrix-assisted laser desorption/ionization, MF = molecular functions, MS = mass spectrometry, PCA = principal component analysis, PPI = protein–protein interaction, RMA = robust multi-array average.

Keywords: bioinformatics analysis, lumbar disc herniation, multichip analysis, the molecular mechanism

1. Introduction

Lumbar disc herniation (LDH) is a group of frequently-occurring and recurring clinical pathologies with neurological symptoms that result from intervertebral disc (IVD) degeneration, annulus rupture, inward movement of the nucleus pulposus under shear stress, which ultimately results in compression of the spinal nerve root.[1] Recent epidemiological studies suggest that the total annual incidence of LDH in the Asian population is 15,877 per 100,000 people (13,181/100,000 in men and 18,588/100,000 women; the sex ratio is male/female 1:1.41). Additionally, every 5 years, the incidence and annual cost of treatment of patients increases by 7.6% and 14.7%, respectively. Reports after age correction suggest that the incidence of LDH increases with age, with a prevalence of 42.6% for people aged 75 to 79. Patients over the age of 65 accounts for 31% of the patient population, and the medical expenses incurred accounts for as much as 40.1% of the total.[2,3]

In a normal IVD, because the nucleus pulposus contains hydrophilic compounds including proteoglycans, the nucleus pulposus is physically stabilized by water. This also generates
certain viscoelasticity that converts the axial load into hoop stress in the IVD. Along with the surrounding dense connective tissue, which ensures structural integrity, the nucleus pulposus provides a unique ability to withstand high pressures, greatly buffering the vertebra, the outer fibrous ring, and the spinal nerve against compressive forces. Repeated long-term wear and other nutritional and environmental factors can lead to gradual degeneration of the IVD and loss of protection by the nucleus pulposus can promote cone compression load, which is transmitted longitudinally to the vertebral endplates and transmitted radially to the surrounding muscle tissues, nerves, and blood vessels. In addition, uneven pressure is applied to the vertebrae and related structural changes can lead to adaptive remodeling of unstable spinal segments, resulting in the formation of osteophyte spurs, which aggravate joint plane imbalance and damage. Chronic nucleus pulposus degradation can also further stimulate hypertrophy and ossification of the spinal ligament, accompanied by capillary hyperplasia and connective tissue infiltration, which promotes a narrowing of the space occupied by the soft tissue of the disc. Chronic ischemia activates a set of unique immune responses, resulting in M1 and M2 macrophage accumulation at the site of compression, further promoting endothelial dysfunction and blood-spinal barrier destruction. At the same time, spinal cord ischemia can increase intramedullary pressure, leading to loss of neurons, demyelination, and further reduction of blood supply caused by the flattening of small blood vessels. A recent view is that the progression of LDH is determined both by environmental and genetic risk factors, and the mechanisms of interaction between different genotypes and environmental factors are also different. Martyrossyan et al. reviewed the association between the development of LDH and the patients’ genetic background and concluded that genetic factors and environmentally-induced genetic changes account for 75% of LDH etiology, especially in inflammatory, degradative, steady-state, and structural systems. Zhang et al. performed a controlled study on 128 LDH patients and 132 age- and sex-matched healthy individuals. Using matrix-assisted laser desorption/ionization (MALDI) and rapid mass spectrometry (MS) techniques, 9 polymorphic loci in 3 genes were analyzed and it was found that the FasL-$844C/T$ (rs763110) and CASP9 -$1263A>G$ (rs4645978) polymorphisms are closely related to LDH, suggesting that they may be sites of potential risk for LDH. Perera et al. evaluated the relationship between single nucleotide variants (SNV) of LDH risk genes and the severity of LDH in patients with chronic lower back pain and found that SNVs located in the aggrecan (ACAN) gene, including rs2272023, rs35430524, rs2882676, rs2351491, rs938609, rs3825994, rs1042630, rs698621, and rs3817428 are closely related to LDH severity. Kurzawski et al. found that the rs676030 polymorphism of the voltage-gated sodium channel Nav1.7 (type IX, alpha subunit, SCN9A) may be associated with the intensity of pain in patients with symptomatic LDH. Sadowska et al. found that interleukin 15 (IL-15) gene expression is associated with age in patients with LDH, interferon-alpha 1 (IFNA1), interleukin 6 (IL-6), IL-15, and transient receptor potential channel-6 (TRPC6) gene expression are closely related to the level of IVD degeneration. In recent years, with the advancement of genome-wide association study (GWAS), IDH disease-related risk factors have been re-identified among known risk genes. These studies not only improve our understanding of the many genetic variations behind the pathophysiological factors underlying IDH, but also help to promote personalized treatment and drug treatment strategies for IDH patients. However, overall, the etiology of IDH is complex and the specific mechanisms have not yet been identified or experimentally verified. Further bioinformatics analysis will help us provide a deeper understanding of the disease and help to further solve multidimensional scientific problems.

2. Methods

2.1. Data acquisition and preprocessing

After searching through the Gene Expression Omnibus (GEO, http://ncbi.nlm.nih.gov/geo/) database, we downloaded raw data from the GSE23130, GSE17077, and GSE15227 chips. The ethics committee not applicable for this study. A total of 57 samples were obtained, of which 35 were from the control group (mildly degraded) and 22 were from the experimental group (severely degraded). All of the chips used tissues derived from IVD tissue from LDH patients and the Thompson classification method was used for tissue classification (Thompson grades IV and V were classified as severely degraded, grades I-III were classified as mildly degraded). At the same time, by matching the platform annotation files corresponding to the 3 sets of chips (GPL1352 [U133 X3P] Affymetrix Human X3P Array; Affymetrix, Santa Clara, CA), we performed gene name conversion on the probe numbers, which helped identify the appropriate genes. Similarly, the original data of the methylated chip GSE129789 matches the GPL21145 Infinium MethylationEPIC platform and contains 8 mildly stage degeneration samples (Thompson grade: I-III) and 8 severely stage degeneration samples (Thompson grade: IV).

During data preprocessing, we used the Affy function to read the original expression value of the chip, robust multi-array average (RMA) was used to normalize the original data, K-nearest neighbor (KNN) was used to fill in missing values and perform other processing. RMA is an algorithm for creating expression matrices from Affymetrix data. The main principles include background correction for original intensity values, Log2 transformation; quantile normalization; perform linear fit on normalized data to obtain expression values for each probe set on each array. The KNN algorithm is one of the classical algorithms commonly used in machine learning. Its principle of dealing with missing values is: when the data of the nth row and the nth column are missing, the nth experiment of the k genes which are similar to the gene m and the nth sample data is replaced. Finally, for the expression matrix of a gene name corresponding to multiple probes, we selected the average value as the gene expression value, and data on anyone probe corresponding to multiple gene names were deleted.

2.2. Differential expressed gene analysis

The Combat function was used to test the heterogeneity of the expression values of the 3 sets of chips, eliminating data offset and batch effects due to different experimental setups. At the same time, principal component analysis (PCA) was used to visualize expression values before and after the application of Combat, and the overall distribution of the data was observed. Finally, we used LIMMA differential expression analysis to evaluate the expression values of the experimental and control groups. Eliminating batch effects can reduce dependence on...
the data itself, reduce the error rate of the analysis, and improve reproducibility; while the limma function is a classical method of differential analysis, based on using the linear model to analyze differences in experimental expression, and is mainly used on high-dimensional data generated by chip microarrays, RNA-seq, quantitative polymerase chain reaction (PCR), and many protein technologies.

Next, using the Benjamini–Hochberg method, we performed the false discovery rate (FDR) correction on the expression data and further estimated the gene expression fold-changes (FC).

We set $\log_2 FC > 2$ and $P < .05$ after FDR correction as a criterion for screening differentially expressed genes.

### 2.3. Analysis of biological functions of differential genes

Using the Database for Annotation, Visualization, and Integrated Discovery Bioinformatics Resources (DAVID database; V6.8; http://david.abcc.ncifcrf.gov/), we further analyzed the Gene Ontology (GO) and Kyoto Encyclopedia of Genes and Genomes (KEGG) pathway information.

The DAVID database is an internationally authoritative functional database of gene annotation, visualization, and integrated exploration that facilitates the integration of genetic pathways and functional enrichment. We applied $P < .05$, which is the generally accepted level at which enrichment is identified, as the cutoff for statistical significance in our GO function and KEGG pathway analyses.

### 2.4. Analysis of protein interaction networks based on differentially expressed genes

After exporting the differentially expressed genes, we performed a protein–protein interaction (PPI) analysis network using the STRING online analysis software (V10.5; http://string-db.org/).

The STRING platform integrates existing experimental conclusions, databases, text mining libraries, gene expression libraries, gene connection data, gene fusion, and gene coexpression databases, and is currently the authoritative tool for protein network prediction. In addition, we also visualized protein interaction networks and key gene screening analyses using Cytoscape software (Institute for Systems Biology, Seattle, Washington 98103, USA) (V3.5.1; http://cytoscape.org/).

### 2.5. DNA methylation data analysis

The GSE129789 methylation original IDAT file using the CHAMP algorithm that processing flow was as follows: the raw data was read; data preprocessing and data quality control; matching with the reference genome; methylation level calculations; differential methylation site and segment analysis; differential gene and CpG island notes. Among them, the proportion of oligonucleotide whose beta value reaction could match the methylated sequence in the methylation analysis was the methylation rate of the sequence; the M value is the log conversion value based on the beta value, which can eliminate interference by the chip probe.

### 3. Results

#### 3.1. Differential gene expression analysis results

Using the analytical methods described in the Methods section, 57 samples (35 from the control group and 22 from the experimental group) from 3 sets of chips were subjected to RMA background correction, normalization, KNN algorithm for missing values, and Combat function to remove batch effects. A total of 149 differentially expressed genes were found to simultaneously satisfy the selection criteria of $\log_2 FC > 2$ (i.e., the difference in expression values is greater than -fold) and the $P$-value after FDR correction is $<.05$. Figure 1 uses PCA analysis to visualize data grouping before and after removal of batch effect during the integration of data from the 3 sets of chips. In Fig. 1A, the distribution of data from the 3 sets of chips is relatively scattered, and the distance between the different sets of chip data is relatively large. Figure 1B shows that the data between the chips is relatively concentrated after batch effects are removed, which is beneficial for subsequent analysis. The heat map in Fig. 1C shows the differences between the top 50 gene expression values, while the difference in color represents the difference in expression values.

### 3.2. Analysis of differential gene biology

Using the DAVID online analysis software, we further analyzed the most significant GO functions and KEGG pathways of differentially expressed genes. Of these, the significant GO functions mainly involved 3 main aspects: biological processes (BP), cellular components (CC), and molecular functions (MF). In terms of BP-related functions, the differentially expressed genes were mainly associated or closely related to GO:0006614—SRP-dependent membrane translation proteins (enriched genes = 13, $P = 5.84E-12$), GO:0000184—nuclear transcriptional mRNA decomposi- tion process (enriched genes = 13, $P = 9.89E-11$), and GO:0006413—translation start (enriched genes = 13, $P = 5.17E-10$). In terms of CC function, the differentially expressed genes are mainly associated with GO:0070062—exosomes (enriched genes = 59, $P = 2.36E-14$), GO:0022625—cytoplasmic large ribosomal subunit (enriched genes = 9, $P = 3.97E-08$), GO:0031012—extracellular matrix (enriched genes = 15, $P = 4.72E-08$). In terms of MF, differentially expressed genes are mainly associated with GO:003735—ribosome structure (enriched genes = 13, $P = 9.49E-08$), GO:0008137—NADH dehydrogenase/ubiquitin activity (enriched genes = 22, $P = 7.96E-05$), and GO:0046961—proton transport ATPase activity, rotation mechanism (enriched genes = 4, $P = 0.005$). Visualization of the GO functions occupied by differentially expressed genes is shown in Fig. 1D.

In terms of their KEGG pathways, the differentially expressed genes were significantly correlated with HSA03010: ribosomal transcriptional translation activity (enriched genes = 13, $P = 8.34E-09$), HSA000 190: oxidative phosphorylation (enriched genes = 12, $P = 7.30E-08$), HSA045 12: Extracellular matrix (ECM) receptor interaction (enriched genes = 10, $P = 1.23E-04$), HSA04060: cytokine–cytokine receptor interaction (enriched genes = 8, $P = 0.0015$), HSA04370: VEGF signaling pathway (enriched genes = 7, $P = 0.041$), HSA01100: metabolic pathway (enriched genes = 20, $P = 0.033$), and HSA04140: regulation of autophagy (enriched genes = 3, $P = 0.048$). The KEGG pathway enrichment results are shown in Fig. 2A.

### 3.3. Analysis of differentially expressed genes and their PPI network

Using the STRING database and Cytoscape software, network diagrams were generated and key genes in the network were screened based on the degree of network connectivity. In this PPI network, ubiquitin A-52 residue ribosomal protein fusion product 1 (UBAS2; ClosenessCentrality: 0.40; Degree: 22), large
Figure 1. Overall data distribution of 3 sets of chips and differential gene expression analysis. A. Presented the PCA is used to show data before and after using the Combat function to remove batch effects. B, Shown the heatmap of differential gene expression: the color of the upper column represents grouping information, the color of the left column represents different gene sets, the color of each cell of the heat map represents its expression value, blue is for low expression, red is for high expression. C, Presented the GO functional analysis results: shapes represent different functional types, circles represent BP, triangles represent CC, and squares represent MF; the size of the shape represents the number of enriched genes; the color represents the statistical value, green represents the control group, and yellow represents the experimental group; the color of each cell of the heat map represents its expression value, blue is for low expression, red is for high expression.

ribosome Protein-P0 (RPLP0), ClosenessCentrality: 0.40; Degree: 18), ribosomal protein L3 (RPL3; ClosenessCentrality: 0.38; Degree: 17), large ribosomal protein P2 (RPL2P2; ClosenessCentrality: 0.37; Degree: 16), and ribosomal protein L27 (RPL27; ClosenessCentrality: 0.40; Degree: 16) have a high degree of connectivity and are considered to be pivotal genes in the LDH differentially expressed gene network. The PPI network is shown in Fig. 2B.
3.4. DNA methylation data analysis

The 450 BeadChip methylation chip contains 96% of the methylation sites of the genome and is the current mainstream chip for methylation analysis (Fig. 3C). Compared with the control group, there were 98 different methylation sites in LDH tissue, 5 of which were downregulated and 93 upregulated. After annotation according to methylation sites, the 3 genes in the present study were differentially methylated and differentially transcribed (including RPL27 and RPLP0) (Fig. 3B and D). We further analyzed and visualized the methylation site and gene expression values corresponding to the RPL27 and RPLP0 gene, and found that the methylation sites of cg12536991 and cg06852319 were decreased in the LDH group and that the RPL27 and RPLP0 transcriptional expression level was upregulated, suggesting transcription of the gene regulated by DNA methylation modification (Fig. 3A and D).

4. Discussion

LDH is a complex spinal disease with high rates of disability, which incurs a heavy medical burden to society and families. In this paper, through multichip integration analysis, IVD samples were divided into the experimental group (Thompson grades IV–V) and the control group (Thompson grades I–III). The original chip data were put through rigorous data processing protocols including background correction and normalization using RMA and KNN algorithms, the Combat function was used to perform the batch correction, and LIMMA was used to screen for differentially expressed genes. Multichip integration analysis revealed that HSA03010: ribosomal transcriptional translation, HSA000 190: oxidative phosphorylation, HSA04512: ECM receptor interaction and other ribosome-related pivotal genes such as UBA52, RPLP0, RPL3, RPLP2, RPL27 are important pathways and regulatory factors in the development of LDH.

Our study suggests that ribosomes are important regulatory mediators of LDH disease progression. However, the specific mechanism is not clear. Kobayashi et al. observed the role of axonal blood flow disorder caused by nerve root compression in lumbar motor neurons by constructing an in vivo model of mechanical compression of nerve roots. In their study, central chromatolysis occurred in motor neurons of the spinal cord 1 week after compression, and ventral horn neuronal apoptosis occurred 3 weeks later. At the same time, the authors observed that ribosomes participate in significant pathological activities during this process. Axonal reactions that may occur with nerve root compression caused by LDH or lumbar spinal stenosis extend to motor neurons in the spinal cord, leading to neuronal chromatolysis and cell death. Similarly, Alamin et al. used real-time PCR to detect targeted 16S ribosomal RNA genes, then used amplicon sequencing to analyze excised disc tissue to find elevated ribosomal gene expression in chronically painful LDH tissues. However, these changes were found to not be related to the pre-study discovery of low toxicity microbial infections. So why do ribosomes occupy such a significant role in the progression of LDH? Recent studies by Slomnicki suggest that in the early stages of neuronal damage, to protect the nucleolus and maintain normal functions, regulatory status in the cell without destroying nucleolar integrity and in response to cell survival and signaling responses to neurotrophins, neurons reduce their ribosome content, resulting in reduced protein synthesis via RNA stress and recruitment of neurotrophins, neurons reduce their ribosome content, resulting in reduced protein synthesis via RNA stress and recruitment of neurotrophins.
proteins are thought to be regulatory sites that govern functional ribosomes, thereby coordinating complex macromolecular motion and translation processes through the “inter-protein communication” connected into a circular network for information transmission and processing.[20] Among the pivotal genes, RPL3 belongs to the ribosome-associated functional large subunit core protein, which interacts with RNA-proteins to fold rRNA into its functionally correct conformation, participates in different functional regions of ribosomes, and is involved in the intercommunication between ribosomes and cytokines. The coding gene UBA52 is not only a ubiquitin donor but also a regulator of the ribosomal protein complex. The UBA52 N-terminal fusion protein containing ubiquitin and C-terminal ribosomal protein L40 (RPL40) are involved in ribosome ubiquitination.[21] However, Artero-Castro et al.[22] believe that RPL0 and RPL2 may not only participate in the activation of reactive oxygen species (ROS) and the corresponding MAPK1/Erk2 signaling pathway, but are also closely related to endoplasmic reticulum stress and autophagy of cells. The ribosomal protein gene RPL27 is well preserved during evolution, relatively conserved and associated with maintaining a specific 3D structure of ribosomal proteins.[22]
In terms of oxidative stress, Xiao et al.[23] believe that oxidative stress plays an important role in the progression of LDH, regulates the regeneration of ECM proteins, and increases the content of nascent collagen and aggrecan in the herniated disc. However, inhibiting the level of tissue oxidative phosphorylation can significantly improve LDH and can promote the recovery of IVD height after LDH. Chang et al.[24] believe that the ECM participates in the formation of the fibrotic matrix in LDH, participates in the three-dimensional structure of the vertebral body and IVD tissue, and changes its corresponding mechanical properties, which may be the root cause of LDH pain. Similarly, studies by Hirose et al.[25] have demonstrated that ECM metabolism and remodeling of the IVD play a crucial role in the etiology and pathogenesis of LDH.

5. Conclusion

Based on multichip integration analysis, we found that, regardless of whether PPI or pathway enrichment analysis was used, ribosome activity or related pivotal genes (including ribosome-related genes such as UBA52, RPLP0, RPL3, RPL2, RPL7)[26] are involved in LDH disease progression, and are closely related to the mechanisms involved in the pathological process of neurogenic damage caused by the disease. Additionally, oxidative stress and ECM interaction pathways are also significantly enriched among the differentially expressed genes of LDH, which may be closely related to aggrecan metabolism, collagen formation, and fibrotic remodeling. However, several limitations have been detected in this study. The findings in this study have still not been experimentally verified and the pathogenesis of LDH is complex. Therefore, specific mechanisms will be further studied in the future.

Author contributions

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