Original Article Transcriptomics data integration and analysis to uncover hallmark genes in hypertrophic cardiomyopathy

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Abstract: Introduction: Hypertrophic cardiomyopathy (HCM) is a heterogeneous disease that mainly affects the myocardium. In the current study, we aim to explore HCM-related hub genes through the analysis of differentially expressed genes (DEGs) between HCM and normal sample groups. Methods: The GSE68316 and GSE36961 expression profiles were obtained from the Gene Expression Omnibus (GEO) database for the identification of DEGs, to explore hub genes, and to perform their expression analysis. Clinical HCM and control tissue samples were taken for expression and promoter methylation validation analysis via RNA-sequencing (RNA-seq) and targeted bisulfite sequencing (bisulfite-seq) analyses. Then, other different bioinformatics tools were employed to perform STRING, IncRNA-miRNA-mRNA regulatory networks, gene enrichment, and drug prediction analyses. Results: In total, the top 20 DEGs, including 10 up-regulated and 10 down-regulated, were obtained from GSE68316. Out of the 20 DEGs, we subsequently identified the 8 most important hub genes including 5 up-regulated genes (EPB42, UQCRH, CA1, PFDN5, and LSM5) and 3 down-regulated genes (RPS24, TNS1, and RPL26). Expression and promoter methylation dysregulation of these genes were further validated on clinical HCM samples paired with controls. Next, we further investigated hub genes' regulatory 6 miRNAs (has-mir-1-3p, has-mir-129-5p, has-mir-16-5p, has-mir-23b-3p, hasmir-27-3p, and has-mir-182-5p) and miRNAs regulatory 4 IncRNAs (NUTMB2-AS1, NEAT1, XIST, and GABPB1-AS1) in this study via the IncRNA-cricRNA-miRNA-mRNA regulatory network. Later on, gene enrichment analysis revealed that hub genes were enriched in various important pathways including Nitrogen metabolism, Ribosome, RNA degradation, Cardiac muscle contraction, and Coronavirus disease, etc. Finally, the drug prediction analysis highlighted different potential candidate drugs for altering the expression of hub genes in the treatment of HCM. Conclusion: In summary, the identification of key hub genes and their enrichment analysis in the current study may shed light on the mechanisms behind the occurrence and development of HCM.

Keywords: Hypertrophic cardiomyopathy, DEGs, hub gene, miRNA

Introduction

Hypertrophic cardiomyopathy (HCM), a complex cardiovascular genetic disorder, is reported in one out of every 500 individuals worldwide [1,

2]. In recent years, awareness of HCM has significantly improved in clinical settings [3, 4]. However, HCM diagnosis is still a complex process, just like other diseases including asthma, mitral regurgitation, and coronary artery disease, resulting in a high mortality rate [5]. From a clinical perspective, the prevalent pathological characteristics of HCM include myocyte hypertrophy, disarray, and interstitial fibrosis [6]. It is important to highlight that approximately 25% of individuals diagnosed with HCM experience left ventricular outflow tract obstruction [7]. The clinical symptoms of HCM vary from patient to patient, due to the natural complex history of the disease [8]. However, if this disease was diagnosed precisely, the patients suffering from HCM could manage this disease more effectively to improve their survival duration.

Currently, echocardiography and some imaging methods such as cardiac resonance imaging are important tools for diagnosing HCM. However, the detection accuracies of these tools are still very much compromised [6]. Moreover, it was also observed that different physical exercise methods have also made great contributions to the management of HCM even without increasing its further risk [9]. The surgical treatment method is a preliminary choice for clinicians to treat HCM. However, due to the increased mortality rate during surgeries, clinicians are still hesitant to choose this method [10].

Recently, a lot of emphasis has been given to research exploring the underlying molecular mechanism of HCM [11, 12]. Genetic testing techniques have now become more accessible at the clinical level. The use of this technique in diagnosing HCM is increasing nowadays [13]. It is important here to understand that HCM has various genotypic and phenotypic variations. HCM was linked with approximately 1400 genetic mutations across more than 10 genes responsible for producing cardiac sarcolemmal protein [2]. Mostly, HCM developed due to a single heterozygous mutation [14, 15]. However, multiple mutations can also lead to a severe form of HCM [16]. According to current research findings, scientists are of the opinion that gene mutations play a significant role in HCM [12]. Various sacromeric genes, namely MYH7, MYBPC3, TPM1, TNNT2, and TNNI3, have been identified as crucial genetic alterations in 34% of individuals with HCM [17-19]. These mutations are typically observed in familial cases with a hereditary pattern. However, it should be noted that gene mutations have not been confirmed in approximately 70% of HCM patients, leaving the underlying genetic causes unclear.

It was also reported in the medical literature that by exploring differentially expressed genes (DEGs) among HCM and normal control samples, potential HCM-associated candidate genes could be figured out [20]. The up-regulation of KRT1 and down-regulation of CYP1A1 at the mRNA level were reported to play a significant role in the development and progression of this disease [21, 22]. Moreover, 3-methylcholanthrene and benzo(a)pyrene-induced cardiac hypertrophy was also associated with the overexpression of CYP1A1 [23].

In this paper, the Gene Expression Omnibus (GEO) datasets related to HCM were analyzed to identify the key hub genes having diagnostic and therapeutic importance for HCM patients. In summary, the results of our study would be helpful to improve the diagnosis and treatment methods for HCM patients.

Methods

Data collection

Normalized gene expression data of HCM patients (n = 7) and normal individuals (n = 5) were obtained from the GSE68316 dataset [24] by using the GEO database (https://www. ncbi.nlm.nih.gov/geo/). Moreover, another GEO dataset GSE36961 consisting of 106 HCM and 39 normal samples was used to obtain the normalized gene expression data for further expression validation. HCM samples in these datasets were collected prior to any clinical intervention.

DEGs identification

For identifying DEGs among HCM and control samples in this study, the R-based limma (Version 3.40.6) package in Bioconductor was utilized [25]. Genes showing *p*-values less than 0.01 and fold changes (FCs) greater than 1.2 were regarded as DEGs. A total of the top 20 genes including 10 up-regulated and 10 down-regulated genes were selected for further study in this manuscript.

Protein-protein interaction (PPI) network construction and hub genes exploration

The PPI network of selected 20 DEGs was constructed via the STRING database [26]. This database was built to provide vital information about predicted and experimental proven protein-protein interactions. In the present study, the PPI of the DEGs having interaction scores greater than 0.7 was considered for further analysis. Subsequently, the top eight hub genes (highly interacting genes) based on the degree method from the obtained PPI network were explored using Cytohubba [27], which is a plugin application in Cytoscape [28].

The IncRNA-miRNA-mRNA regulatory network

To develop the potential IncRNA-miRNA-mRNA regulatory network of the hub genes we used the following different online databases. The miRNAs were predicted via miRTarBase (http:// mirtarbase.mbc.nctu.edu.tw/) [29], and miRDB (http://www.mirdb.org) [30] databases. The MiRcode (http://www.mircode.org/) repository [31] was utilized to predict IncRNAs targeting miRNAs. Ultimately, the IncRNA-miRNA-mRNA regulatory network was incorporated using Cytoscape.

Enrichment analysis

Gene Ontology (GO), BP (biological processes), CC (cellular components), MF (molecular function), and Kyoto Encyclopedia of Genes and Genomes (KEGG) enrichment analyses of the hub genes were carried out using DAVID (https://david.ncifcrf.gov/). This is a famous online enrichment annotation tool to explore the biological significance of a given list of genes [32, 33].

Drug prediction analysis

DrugBank (http://www.drugbank.ca) database [34], which contains around seven thousand drug entries and four thousand proteins data, was used in this study to evaluate hub genesassociated potential targeted drugs.

Clinical HCM and normal control tissue sample collection

Following the approval of the ethics committee, we conducted a prospective collection of 25 pairs of HCM tissue and corresponding normal samples from patients who visited the Institute of Nishtar Hospital, Multan between August 2022 and May 2023. Prior to their participation, all individuals provided informed consent by signing consent forms. All patients included in the study were diagnosed with HCM and had not undergone any type of therapy.

Total RNA, DNA extraction, RNA sequencing (RNA-seq) and targeted bisulfite sequencing (targeted bisulfite-seq) analyses

Total RNA extraction from all these HCM and control tissues was done using the TRIzol[®] reagent method [35], and DNA extraction was done via the organic method [36-38]. RNA and DNA samples were sent to the Beijing Genomics Institute (BGI) Company for RNA-sequencing and targeted bisulfite-sequencing analysis.

Following RNA-seq and targeted bisulfite-seq analyses, the gene expression values of the hub genes were normalized using reads per kilo base million reads (RPKM) and fragments per kilo base million reads (FPKM). While methylation values were normalized as beta values. The obtained FPKM and beta values against hub genes in HCM and normal control samples were compared to identify differences in the expression and methylation levels.

Statistics of bioinformatics analysis

DEGs were identified using a t-test [39, 40]. For GO and KEGG enrichment analysis, we used Fisher's Exact test for computing statistical difference [41]. All the analyses were carried out in R version 3.6.3 software.

Results

The overall study design is given in Figure 1.

DEGs and hub genes between HCM patients and normal individuals

The variations in the gene expression pattern between normal individuals and patient group in disease may be closely associated with the occurrence of that disease [42]. The normalized gene expression data were taken from the GSE68316 dataset via the GEO database and analyzed to discover DEGs specifically associated with HCM. By selecting *p*-values less than 0.01 and FC greater than 1.2 as the screening thresholds, a total of 850 DEGs were explored between HCM patients and healthy controls (Figure 2A). From the obtained DEGs, there were a total of 636 genes that were up-regulated and 270 genes that were down-regulated in HCM samples relative to healthy controls (Figure 2B). We selected the top 20 DEGs



Figure 1. The overall workflow sheet diagram of our study.

including 10 up-regulated and 10 down-regulated genes for further study in this manuscript (**Table 1**). After constructing a PPI of these 20 DEGs (**Figure 2C**), we further shortlisted the eight most important genes as the hub genes including Erythrocyte Membrane Protein Band 42 (EPB42), Ribosomal Protein S24 (RPS24), Ubiquinol-Cytochrome C Reductase Hinge Protein (UQCRH), Tensin 1 (TNS1), Ribosomal Protein L26 (RPL26), Carbonic Anhydrase 1 (CA1), Prefoldin Subunit 5 (PFDN5), and U6 SnRNA-Associated Sm-Like Protein (LSM5) (**Figure 2D**).

Validation of hub genes using additional GEO dataset

To validate the expression levels of hub genes in another GEO expression dataset (GSE36961), the normalized gene expression data from 106 HCM patients and 39 healthy samples were obtained from the GEO database. Expression analysis of the hub genes using GSE36961 revealed that EPB42, UQCRH, CA1, PFDN5, and LSM5 hub genes were significantly up-regulated while RPS24, TNS1, and RPL26 hub genes were significantly down-regulated in HCM samples relative to healthy controls (**Figures 3** and **4**). Therefore, the re-analysis of hub gene expression further proves the important roles of these genes in HCM occurrence and progression. In summary, it is suggested that these hub genes between HCM and healthy individuals may be aberrantly expressed genes related to HCM.

IncRNA-miRNA-mRNA regulatory network analysis

We constructed a IncRNAmiRNA-mRNA regulatory network of the hub genes using different online databases (referred to methodology) in the current study. We noticed in the current study that there were 189 miRNAs that target all hub genes including EPB42, RPS24, UQCRH, TNS1, RPL26, CA1, PFDN5, and LSM5 (**Figure 5A**). Further

interaction analysis showed that there were 8 miRNAs (has-mir-23b-3p, has-mir-124-3p, has-mir-129-5p, has-mir-182-5p, has-mir-16-5p, has-mir-1-3p, has-mir-7b-5p, and has-mir-27a-23p) in total predicted 189 miRNAs which predicted to be targeting all the hub genes collectively (**Figure 5B**). Additional analysis revealed that, out of the 8 miRNAs, 6 miRNAs (has-mir-1-3p, has-mir-129-5p, has-mir-16-5p, has-mir-23b-3p, has-mir-27-3p, and has-mir-182-5p) were also found targeted by 246 IncRNAs (**Figure 6A**) and out of which only 4 IncRNAs including NUTMB2-AS1, NEAT1, XIST, and GABPB1-AS1 revealed to be collectively targeting all 6 miRNAs (**Figure 6B**).

GO enrichment analysis

GO analysis for predicting BP, CC, and MF of the hub genes was conducted via the DAVID tool. The identified hub genes were highly enriched in Reg. of DNA damage response, signal transduction, by p53 class mediator resulting, Pos. reg. of intrinsic apoptotic signaling pathway by p53 class mediator, Pos. reg. of intrinsic apoptotic signaling pathway in response to DNA damage, SRP-dependent cotranslational pro-



Figure 2. This figure illustrates key aspects of the analysis, incorporating visual representations of differentially expressed genes (DEGs) and hub genes. (A) showcases a volcano graph depicting the DEGs identified in the GSE68316 dataset. (B) presents an overarching heatmap displaying the expression patterns of DEGs identified in the GSE133054 dataset. (C and D) depict protein-protein interaction (PPI) networks for the DEGs and hub genes, respectively.

tein targeting to membrane, Protein targeting to ER, Establishment of protein localization to endoplasmic reticulum, Erythrocyte homeostasis, Nuclear-transcribed mRNA catabolic processes, nonsense-mediated decay, and Protein targeting to endoplasmic reticulum, etc. BP terms (**Figure 7A**).

Concerning CC, the identified hub genes were mainly enriched in Lsm-7-Pat complex, Prefoldin complex, Lsm2-8 complex, Mitochondrial respiratory chain complex III, Respiratory chain complex III, U6 snRNP, Cytosolic ribosome, Ribosomal subunit, Ribosome, Ribonucleoprotein complex, etc. CC terms (**Figure 7B**). Regarding MF, the identified hub genes were mainly involved in Arylestrease activity, Ubiquinol-cytochrome-c reductase activity, Oxidoreductase activity, acting on diphenois and related substances as donors, Protei glutamine gamma-glutamyltransferase activity, Carbonate dehydratase activity, Structural constituent of ribosome, etc. MF terms (**Figure 7C**).

KEGG pathway analysis

To further clarify the roles of hub genes (EPB42, RPS24, UQCRH, TNS1, RPL26, CA1, PFDN5, and LSM5) in the development of HCM, the KEGG pathway analysis was also performed by

Sr. No	Gene symbol	Expression Status	Log FC	Adjust <i>P</i> -value
1	SELENBP1	Up-regulation	6.481	2.22E-17
2	LSM5	Up-regulation	6.360	2.91E-13
3	HBG1	Up-regulation	5.848	3.36E-06
4	TROAP	Up-regulation	5.734	6.80E-17
5	BTN3A2	Up-regulation	5.246	4.26E-17
6	UQCRH	Up-regulation	5.117	1.72E-18
7	CA1	Up-regulation	5.066	1.24E-13
8	SLCA4A1	Up-regulation	4.797	8.42E-15
9	PFDN5	Up-regulation	4.557	2.08E-18
10	PDE4DIP	Up-regulation	4.386	7.35E-19
11	COMMD6	Down-regulation	3.843	2.02E-19
12	P2RY14	Down-regulation	3.299	7.18E-13
13	CCDC7	Down-regulation	3.080	4.90E-17
14	KIAA1841	Down-regulation	3.010	1.26E-19
15	TNS1	Down-regulation	2.997	2.31E-17
15	RPL26	Down-regulation	2.995	6.68E-20
17	RPS24	Down-regulation	2.902	6.91E-19
18	FURIN	Down-regulation	2.793	1.51E-13
19	EPB42	Down-regulation	2.764	5.80E-17
20	HBM	Down-regulation	2.682	2.17E-17

Table 1. The top 20 dysregulated genes from the GSE68316 dataset

DAVID. The selected hub genes were mainly involved in Nitrogen metabolism, Ribosome, RNA degradation, Cardiac muscle contraction, and Coronavirus disease, etc. KEGG terms (Figure 8A-C).

Drug prediction analysis

Medical treatment is the preliminary choice to handle disease for patients who are suffering from HCM. Therefore, a selection of suitable candidate potential drugs is necessary. In the current study, with respect to identified hub genes, we explored some suitable therapeutic drugs for the treatment of HCM via the DrugBank database. For example, Rofecoxib drug was identified as the negative expression regulator of EPB42 mRNA expression (**Table 2**) while Acetaminophen was identified as the negative expression regulator of RPS24 mRNA expression (**Table 2**).

Hub genes expression and promoter methylation analysis using clinical HCM and control samples

In the current study, using RNA-seq and targeted bisulfite-seq data of 25 HCM and control

samples, the expression and promoter methylation levels of identified hub genes were validated. The expression and promoter methylation levels of these genes were validated using FPKM and beta values, which are quantitative values with widespread use in the RNA-seq and targeted bisulfite-seq analyses.

As shown in **Figure 8A**, it was noticed that EPB42, RPS24, UQCRH, TNS1, RPL26, CA1, PFDN5, and LSM5 hub genes were expressed in both HCM and normal control samples and FPKM values of EPB42, UQCRH, CA1, PFDN5, and LSM5 were higher while FPKM values of RPS24, TNS1, and RPL26 were higher in HCM samples as compared to normal controls (**Figure 9A**). Moreover, the beta values of the EPB42, RPS24, UQCRH, TNS1, RPL26, CA1, and PFDN5 were lower, while beta values of RPS24, TNS1, and RPL26 were higher in HCM samples as compared to normal controls (**Figure 9A**).

Discussion

Exploring the exact underlying molecular mechanisms of HCM can help to improve the clinical management of this disease. The current study



Figure 3. The mRNA expression levels of the shortlisted 4 hub genes were validated using the GSE36961 dataset. (A) EPB42, (B) RPS24, (C) UQCRH, and (D) TNS1. A P < 0.01 was regarded as the selection criteria. EPB42 = Erythrocyte Membrane Protein Band 42, RPS24 = Ribosomal protein S24, UQCRH = Ubiquinol-cytochrome c reductase hinge protein, TNS1 = Tensin 1.



Figure 4. The mRNA expression levels of the shortlisted 4 hub genes were validated using the GSE36961 dataset. (A) RPL25, (B) CA1, (C) PFDN5, and (D) LSM5. A P < 0.01 was regarded as the selection criteria. RPL25 = Ribosomal Protein L26, CA1 = CA1 carbonic anhydrase 1, PFDN5 = Prefoldin subunit 5, LSM5 = U6 SnRNA-Associated Sm-Like Protein.

A												
	hsa-mir-1301-3p	hsa-mir-34b-5p	hsa-mir-1304-5p	hsa-mir-378c	hsa-mir-4326	hsa-mir-423-3p	hsa-mir-628-5p	hsa-mir-20a-3p	EPB42	hsa-mir-578	hsa-mir-450a-5p	hsa-mir-142-5p
	hsa-mir-217	hsa-mir-15b-5p	hsa-mir-424-5p	hsa-mir-148a-3p	hsa-mir-26a-5p	hsa-mir-100-5p	hsa-mir-940	hsa-mir-652-3p	hsa-mir-93-3p	hsa-mir-223-5p	hsa-mir-191-5p	hsa-mir-618
	hsa-mir-609	hsa-mir-21-5p	hsa-mir-1910-5p	hsa-mir-588	hsa-mir-4265	hsa-mir-5010-3p	hsa-mir-1-3p	hsa-mir-205-5p	hsa-mir-141-5p	hsa-let-7a-5p	hsa-mir-33a-5p	hsa-mir-924
	hsa-mir-9-5p	CA1	hsa-mir-7-1-3p	hsa-mir-135b-5p	hsa-mir-376a-5p	hsa-mir-342-3p	hsa-mir-378d	hsa-mir-3616-3p	hsa-let-7c-5p	hsa-mir-195-5p	hsa-let-7b-5p	hsa-mir-145-3p
	hsa-mir-2116-5p	hsa-mir-26b-5p	hsa-mir-148a-5p	hsa-mir-342-5p	hsa-mir-23b-5p	hsa-mir-185-5p	hsa-mir-628-3p	hsa-mir-98-5p	hsa-mir-140-3p	hsa-mir-130a-3p	hsa-mir-99a-5p	hsa-mir-33a-3p
	hsa-mir-1248	hsa-mir-3152-3p	hsa-mir-542-3p	hsa-mir-151a-5p	hsa-mir-654-3p	hsa-let-7g-5p	hsa-mir-129-2-3p	hsa-mir-200c-5p	hsa-mir-335-5p	hsa-mir-3679-3p	hsa-mir-28-5p	hsa-let-7f-5p
	PFDN5	hsa-mir-15a-5p	hsa-mir-424-3p	hsa-mir-942-5p	hsa-mir-200a-5p	hsa-mir-338-3p	hsa-mir-128-3p	hsa-mir-4317	hsa-mir-122-5p	hsa-mir-3198	hsa-mir-574-5p	hsa-mir-1224-5p
	hsa-mir-34a-5p	hsa-mir-877-3p	hsa-let-7e-5p	hsa-mir-186-3p	hsa-mir-124-3p	hsa-mir-126-3p	hsa-mir-4446-5p	hsa-mir-7-5p	hsa-mir-659-3p	hsa-mir-380-3p	hsa-mir-21-3p	hsa-mir-877-5p
	hsa-mir-193b-3p	hsa-mir-143-3p	hsa-mir-454-3p	TNS1	LSM5	hsa-mir-608	hsa-mir-3125	hsa-mir-25-3p	hsa-mir-107	hsa-mir-483-3p	hsa-mir-4279	hsa-mir-19b-1-5p
	hsa-mir-147a	hsa-mir-92a-3p	hsa-mir-17-5p	hsa-mir-7974	hsa-mir-130b-5p	hsa-mir-92a-1-5p	hsa-mir-16-5p	hsa-mir-484	hsa-mir-378a-3p	hsa-let-7i-5p	hsa-mir-132-3p	hsa-mir-148b-3p
	hsa-mir-642a-5p	hsa-mir-10b-3p	hsa-mir-345-5p	hsa-mir-1237-3p	hsa-mir-615-5p	hsa-mir-23b-3p	hsa-mir-27b-3p	hsa-mir-431-5p	hsa-mir-671-5p	hsa-mir-590-3p	hsa-mir-511-5p	hsa-mir-96-5p
	hsa-mir-101-3p	hsa-mir-935	hsa-mir-127-5p	hsa-mir-409-3p	hsa-mir-33b-5p	hsa-mir-361-3p	RPL26	hsa-mir-216a-3p	hsa-mir-1343-3p	hsa-mir-340-5p	hsa-mir-375	hsa-mir-22-5p
	UQCRH	hsa-mir-221-3p	hsa-mir-135a-5p	hsa-mir-29c-3p	hsa-mir-502-5p	hsa-mir-874-3p	hsa-mir-103a-3p	hsa-mir-182-5p	RPS24	hsa-mir-129-5p	hsa-mir-204-5p	hsa-mir-155-5p
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Figure 5. PPI networks highlighting associations between miRNAs and identified hub genes. A. A network of overall predicted miRNAs targeting hub genes, and B. A PPI network between meaningful 8 miRNAs and hub genes. The red nodes are hub genes, green nodes are the miRNAs, and blue nodes are the meaningful miRNAs. PPI = Protein-protein interaction.

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G2E3-AS1	KTN1-AS1	CBR3-AS1	DHRS4-AS1	LINC01876	C9orf163	AP4B1-AS1	PCGEM1	RASAL2-AS1	BASP1-AS1	SLFNL1-AS1	LINC02478	LINC01089	XIST	HOTAIRM1	ASH1L-AS1
ACTN1-AS	1 SNHG17	hsa-mir-1-3p	PAX8-AS1	PWAR5	ROSER2-AS	LINC01518	LINC01247	FER1L6-AS2	AMTOR5-AS	LINC01197	SLC25A25-AS	X1B-SULT1	CCDC18-AS1	SMIM25	GPR158-AS1
LINC0026		LINC02128	MEG8	LINC00622	KCNMA1-AS1	DLEU2	MEM147-AS		MCX5-GPRA	S HOXC-AS3	CDKN2B-AS1	NRSN2-AS1	LINC01441	THRB-IT1	LINC00511
LINC0112	TMEM132D-A	S LINC01547	FAM198B-AS	GAS5	PCAT1	ZNF213-AS1	NEAT1	EBLN3P	DLX6-AS1	N4BP2L2-IT2	ZNF337-AS1	NBR2	AFAP1-AS1	PSMA3-AS1	MBNL1-AS1
LIPE-AS1	LINC00922	LINC01550	AGAP2-AS1	NORAD	RGPD4-AS1	LINC01001	RNF139-AS1	C1RL-AS1	hsa-mir-23b-3	LINC02434	LINC01963	PRR7-AS1	LINC00943	LINC01579	SCAMP1-AS1
LINC0113	TSPEAR-AS	MUC20-OT1	LBX1-AS1	PB41L4A-AS	LINC01355	RMRP	MIAT	TMEM202-AS	RASSF8-AS1	TEX41	FAM239A	NUTM2A-AS	LINC00324	FAM225A	AZIN1-AS1
DLEU2L	CASC9	LINC00963	GSN-AS1	MIR600HG	COX10-AS1	BOLA3-AS1	HOTAIR	RMST	UCA1	TUG1	LINC00997	LINC00649	SCGB1B2P	TNRC6C-AS1	STX17-AS1
hsa-mir-16-	LINC02245	ZSCAN16-AS	1 <mark>ZNF674-AS</mark> 1	PCBP1-AS1	sa-mir-129-5	LINC00242	LINC01572	LINC01087	FGD5-AS1	SENCR	MIR503HG	OTUD6B-AS1	SNHG16	VASH1-AS1	MEM161B-AS1
ERVK13-1	CERS6-AS1	RUNDC3A-AS	GACAT2	LINC00662	LINC01534	NUTM2B-AS1	SNHG12	LINC01485	FAM225B	CKMT2-AS1	LINC01703	1D3P1-DHX4	0PC8orf49	ZNF561-AS1	STX18-AS1
PRKCQ-AS	1 FAM239B	TTN-AS1	ZMIZ1-AS1	UBA6-AS1	LINC00173	HCG18	RAF3IP2-AS	ATXN8OS	L5P-PVRIG2	LINC00639	LINC02381	HELLPAR	SDCBP2-AS1	TRIM52-AS1	SNHG14
LINC0157	ECDC144NL-A	ZNF460-AS1	MIR4697HG	LINC00665	LINC01184	TOB1-AS1	OLMALINC	IQCH-AS1	SND1-IT1	LINC00891	ENTPD1-AS1	FRMD6-AS1	MIR497HG	SOX9-AS1	LINC00667
BMS1P4	PARD6G-AS	LINC01727	MIR4453HG	LINC00909	SLC9A3-AS1	SNHG5	SNHG3	MIR29B2CH0	NFRSF14-AS	UBL7-AS1	DNMBP-AS1	EXTL3-AS1	RPARP-AS1	LINC01515	ST7-AS1
MIR4458H	GMIR4435-2H	LINC00174	MCM3AP-AS	1 ACAP2-IT1	LINC00852	CYTOR	OIP5-AS1	MIR100HG	LINC01122	GUSBP11	TRPM2-AS	SNHG10	MALAT		00894
LINC0203	ERI3-IT1	MSC-AS1	TRG-AS1	SNHG22	PEG13	VPS9D1-AS1	MZF1-AS1	LINC00472	ZEB1-AS1	DLG1-AS1	FENDRR	TTC28-AS1	TUSC7	PITPNA-AS1	MAPT-IT1
LINC0063	HAGLR	SNHG25	ZSWIM8-AS1	HCP5	ST7-OT4	MALINC1	CASC19	LINC01227	MAGI2-AS3	LINC01747	PTPRG-AS1	hsa-mir-27a-3	ARHGAP11B	LINC01278	C18orf15
DNAJC27-A	51 LINC00473	SNHG1	LINC00689	GAS1RR	LINC00707	HCG17	MEG3	LINC00843	LINC00641	LC25A21-AS	LINC01422	PVT1	SH3BP5-AS1	GABPB1-AS1	LINC01995
TYMSOS	TBX2-AS1	DANT2	PSMD6-AS2	HUMPD3-AS	1 FAM30A	LINC-PINT	SP2-AS1	LINC00482	LINC01250	MCPH1-AS1	LINC01018	NPTN-IT1	nsa-mir-182-5	p	



Figure 6. PPI networks highlighting associations between IncRNAs and identified miRNAs. (A) A network of overall predicted IncRNAs targeting miRNAs, and (B) A PPI network between meaningful 4 IncRNAs and miRNAs. Pink nodes are IncRNAs and blue nodes are the miRNAs. PPI = Protein-protein interaction.

was based on DEGs in HCM patients relative to healthy individuals, combined with protein-protein interaction analysis to discover a few key hub genes associated with HCM. We analyzed HCM GSE68316 and GSE36961 datasets from the GEO database to identify the top 20 DEGs, hub genes, and explore the molecular mechanisms of this disease. By constructing a PPI network of the identified DEGs, we found 8 hub genes in HCM patients, namely EPB42, RPS24, UQCRH, TNS1, RPL26, CA1, PFDN5, and LSM5. The expression verification of these hub genes via the HCM GSE36961 dataset revealed that EPB42, UQCRH, CA1, PFDN5, and LSM5 hub genes were significantly up-regulated while RPS24,



Figure 7. GO enrichment analysis of the hub genes via the DAVID. (A) BP terms, (B) CC terms, and (C) MF terms. A P < 0.01 was regarded as the selection criteria. GO = Gene Ontology, BP = Biological process, CC = Cellular components, MF = Molecular function.



Figure 8. KEGG enrichment analysis of the hub genes via the DAVID. (A) KEGG terms, (B) KEGG terms phylogram, and (C) Most significant KEGG term. A P < 0.01 was regarded as the selection criteria. KEGG = Kyoto Encyclopedia of Genes and Genomes.

Sr. No	Hub gene	Drug name	Effect	Reference	Group
1	EPB42	Rofecoxib	Increased expression of EPB42 mRNA	A22107	Approved
2	RPS24	Acetaminophen	Decreased expression of RPS24 mRNA	A20420	Approved
		Vincristine		A22706	
3	UQCRH	Disulfiram	Decreased expression of UQCRH mRNA	A21750	Approved
4	TNS1	Ciglitazone	Increased expression of TNS1 mRNA	A21647	Approved
		Resveratrol		A24017	
5	RPL26	Bortezomib	Increased expression of RPL26 mRNA	A21434	Approved
6	CA1	Rifampicin	Decreased expression of RPS24 mRNA	A24040	Approved
7	PFDN5	Estradiol	Decreased expression of PFDN5 mRNA	A21098	Approved
		Valproic acid		A24690	
8	LSM5	Tretinoin	Decreased expression of LSM5 mRNA	A24454	Approved

Table 2. Hub genes associated drugs from DrugBank database

TNS1, and RPL26 hub genes were significantly down-regulated in HCM samples relative to healthy controls.

EPB42 is an ATP-binding protein, which is reported to regulate the association between protein 3 and ankyrin [43]. This protein may also have a key role in maintaining erythrocyte shape and providing mechanical properties [43]. Mutations in this gene or its dysregulation are linked to the recessive spherocytic elliptocytosis [44]. However, we are the first to report the dysregulation of this gene in HCM. RPS24 and RPL26 genes encode ribosomal proteins, which are part of the ribosome machinery [45]. Ribosomes are the organelles in the cells which carry out the synthesis of proteins [46]. Mutations in these genes are known to be associated with Diamond Blackfan Anemia and HCM [47]. UOCRH protein plays an important role to affect the mitochondrial function [48]. Recent data suggested that this protein also plays an important role in cardioprotection and its dysregulation may disturb normal heart functioning [49]. The TNS1 gene encodes for the TNS1 protein, which plays a key role in attaching the membrane of a cell to the extracellular matrix [50]. In recent research, the dysregulation of TNS1 resulted in myocardial infraction in zebra fish model [50]. CA1 gene in humans encodes for carbonic anhydrase 1 enzyme, which catalyzes the conversion of carbon dioxide to carbonic acid [51]. The dysregulation of this gene was previously correlated with heart failure and other diseases [52]. PFDN5 gene codes for a prefoldin alpha subunit family member involved in the stabilization of newly synthesized polypeptide [53]. Dysregulation of the PFDN5 gene was earlier linked to the development and progression of cancer [54]. However, we are the first to report the dysregulation of this gene in HCM. LSM5 encodes a protein that plays a key role in the assembly of spliceosome [55]. Abnormal expression of this gene was associated with circadian rhythms and HCM [56].

In recent times, the participation of miRNAs, and IncRNAs in HCM development is a hot topic for researchers. In this regard, different studies have discovered the involvement of a variety of miRNAs and IncRNAs in HCM. For example, the IncRNA Rp5-833A20.1 was involved in the prevention of miR-382-5p to target NFIA, which is associated with HCM [57]. The IncRNA MIAT was verified to promote myocardial fibrosis after acute ischemia by targeting miRNA24 to target two important mRNAs including Furin and TGF-B1 [58]. The IncRNA ROR is also reported to sponge miR-133 to dysregulate two important mRNAs (ANP and BNP), associated with HCM [59]. In the current manuscript, concerning IncRNA-miRNA-mRNA regulatory network, we explored 6 miRNAs (has-mir-1-3p, has-mir-129-5p, has-mir-16-5p, has-mir-23b-3p, has-mir-27-3p, and has-mir-182-5p) targeting all the shortlisted hub genes and getting targeted by one only 4 IncRNAs including NUTMB2-AS1, NEAT1, XIST, and GABPB1-AS1. These critical data may expand our knowledge about the regulatory mechanisms of HCM. In addition, functional analysis revealed that HCM-associated dysregulated hub genes were majorly involved in a variety of GO and KEGG terms (Figures 6 and 7). Lastly, we have also suggested a few hub gene-related therapeutic



Figure 9. Validating hub gene expression and promoter methylation using clinical HCM and control samples via RNA sequencing and targeted bisulfite sequencing analysis. (A) FPKM values based expression validation of the hub genes, and (B) Beta values based promoter methylation validation of the hub genes. HCM = Hypertrophic cardiomyopathy, FPKM = Fragments Per Kilobase of transcript per Million mapped reads.

drugs from the DrugBank database whose use might help to treat HCM patients.

Previously, it was reported in the medical literature that HCM is the outcome of genetic mutations across the oncoprotein coding gene in Mendel's autosomal dominant genetic pattern [60]. The discovery of novel dysregulated hub genes in the current study may be helpful for clinically determining the diagnosis of HCM patients. The outcomes of the present study may also guide clinicians to design and use novel treatment therapies for HCM patients.

Conclusion

In the present study, 8 key dysfunctional hub genes (EPB42, RPS24, UQCRH, TNS1, RPL26,

CA1, PFDN5, and LSM5) were discovered via detailed analysis between HCM and control samples. Via further enrichment analysis, we revealed that underlying molecular mechanisms behind HCM were associated with Nitrogen metabolism, Ribosome, RNA degradation, Cardiac muscle contraction, and Coronavirus disease signaling pathways. However, this study requires further experimentation to validate the analysis results. One limitation of this study is the focus on developmental diseases as the research subject. As a result, there is no availability of clinical data to accurately determine the patients' origins, ages, and genders. This restricts the generalizability and applicability of the findings to specific groups. Additionally, there is a scarcity of clinical samples to validate crucial analysis results.

Disclosure of conflict of interest

None.

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