



Comprehensive RNA-Seq Data Analysis Identifies Key mRNAs and IncRNAs in Atrial Fibrillation

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Long non-coding RNAs (InCRNAs) are an emerging class of RNA species that may play a critical regulatory role in gene expression. However, the association between InCRNAs and atrial fibrillation (AF) is still not fully understood. In this study, we used RNA sequencing data to identify and quantify the both protein coding genes (PCGs) and InCRNAs. The high enrichment of these up-regulated genes in biological functions concerning response to virus and inflammatory response suggested that chronic viral infection may lead to activated inflammatory pathways, thereby alter the electrophysiology, structure, and autonomic remodeling of the atria. In contrast, the downregulated GO terms were related to the response to saccharides. To identify key InCRNAs involved in AF, we predicted InCRNAs regulating expression of the adjacent PCGs, and characterized biological function of the dysregulated InCRNAs. We found that two InCRNAs, ETF1P2, and AP001053.11, could interact with protein-coding genes (PCGs), which were implicated in AF. In conclusion, we identified key PCGs and InCRNAs, which may be implicated in AF, which not only improves our understanding of the roles of InCRNAs in AF, but also provides potentially functional InCRNAs for AF researchers.

Keywords: long non-coding RNAs, atrial fibrillation, RNA-Seq, genes, protein coding genes

INTRODUCTION

Atrial fibrillation (AF), one of the most common serious arrhythmia worldwide, whose extreme complications such as heart failure and embolic stroke are often of high risks and associated with increasing morbidity and mortality (Conen et al., 2011). Atrial remodeling, both electrical and structural, are important characteristics in AF (Li et al., 2017; Allessie et al., 2002). AF could bring permanent changes such as enlarged left and right atrial size. Moreover, increasing left atrial volume has been stated as a risk factor of cardioembolic stroke, and it is critical to interpret the mechanism behind this to improve the stroke prevention strategy.

The etiology of AF has not been fully elucidated as a varying range of factors would contribute to AF, such as family history, unhealthy lifestyle, high blood pressure and other diseases (Shi et al., 2013). With the development of next-generation sequencing technologies, non-coding

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RNAs (ncRNAs) emerged as the epicenter for researchers to further explore the genetic cause behind AF. ncRNAs, which can be subdivided into small ncRNAs (< 200 nt) and long ncRNAs (lncRNAs), are not translated in proteins, but some of them are capable of regulating various cellular processes such as the expression of certain genes. Evidences have verified that many lncRNAs, often generated from transcriptional units, play a critical role in several cardiovascular diseases (Su et al., 2018), and it is of great importance to survey how they function in AF and how they are connected with atrial remodeling. Several researches are conducted to explore how lncRNAs acted as regulators in atrial electrical remodeling, revealing that TCONS_00075467 may help decrease AF vulnerability through suppressing the electrical remodeling (Li et al., 2017). Recent reports have also unveiled that lncRNAs can act as modulators of miRNA levels in various cardiac diseases (Greco et al., 2018). Also, inflammation and AF are confirmed to have a close relationship. Abundant inflammatory markers and higher ratios of neutrophil and lymphocyte are often observed in patients with AF (Hu et al., 2015), and AF subsequently triggers more inflammatory response, which in turns results in worsened conditions. Exploring active lncRNAs in inflammation would shed light on the prevention, diagnosis, and therapeutic strategies of AF, and help elucidate underlying mechanisms.

In the present study, we identified differentially expressed lncRNAs and mRNAs in patients with AF and predicted lncRNA function in a co-expression-based manner. Prediction of cis-acting lncRNAs and functional annotation of dysregulated lncRNAs screened out some critical lncRNAs implicated in AF, which not only improves our understanding of the roles of lncRNAs in AF, but also provides potentially functional lncRNAs for AF researchers. In addition, as evidence proves that a variety of inflammation-associated cytokines and chemokines are involved in the pathogenesis of AF (Schnabel et al., 2009), we further investigate whether our findings are related to cytokines and chemokines in certain aspects.

MATERIALS AND METHODS

Data Collection

We collected RNA sequencing data of 6 cases with AF and 6 controls from Sequence Read Archive (SRA, https://www. ncbi.nlm.nih.gov/sra) database (Leinonen et al., 2011) with an accession number SRP093226, which was provided by previous study (Yu et al., 2017). We uncompressed the SRA files by fastq-dump with the option *–split-files*, which generated two paired fastq files.

Read Mapping and Gene Expression Quantification

For each sample, we first mapped the RNA-seq reads to UCSC hg19 human reference genome (www.genome.ucsc.edu) using hisat2 (Kim et al., 2015), and then sorted the SAM files by samtools. With the gene annotation from GENCODE v19 (Harrow et al., 2012), the gene expression was estimated by

the StringTie (Pertea et al., 2015) and ballgown pipeline. We considered genes with biotypes, including 'processed_transcript', 'pseudogene', 'lincRNA', '3prime_overlapping_ncrna', 'antisense', 'sense_intronic', and 'sense_overlapping', as lncRNAs.

Differential Expression Analysis

The FPKM-based expression was used to identify differentially expressed genes. The gene expression values were first transformed to \log_2 (FPKM + 1), and then tested for differential expression by t test. The differentially expressed genes were identified with the thresholds of *P*-value <0.05 and fold change >2 or <1/2.

Gene Ontology Enrichment Analysis

The Gene Ontology (GO) enrichment analysis was implemented in R with clusterProfiler package (Yu et al., 2012), which used overrepresentation enrichment analysis (ORA) to identify enriched GO terms. The GO terms were deemed to be significantly enriched if the adjusted P < 0.05 and the gene count in each GO term was more than 3.

Functional Annotation of IncRNAs

The biological function of lncRNAs was annotated by overrepresentation enrichment analysis (ORA) of co-expressed protein-coding genes (PCGs). The PCGs were defined as co-expressed genes with a given lncRNA if the P < 0.0001 for the correlation coefficient test.

Identification of Cis-Acting IncRNAs

As the lncRNAs may regulate the expression levels of the corresponding adjacent PCGs by cis-acting manner, the cisacting lncRNAs were identified if the lncRNA and its adjacent PCGs (within one million base pairs) exhibit highly correlated expression (Pearson correlation coefficient > 0.5 or < -0.5).

RESULTS

RNA Sequencing Method Reveals Diverse RNAs in Both AF and Control Groups

The analysis of sequencing data with 6 AF patients and 6 controls allowed us to identify 15,147 genes in total (FPKM > 1 in at least one sample), which consisted of 29 RNA categories, including protein-coding genes (PCGs), pseudogenes, antisense RNAs and etc. (**Figure 1A**). The PCGs, pseudogenes, and antisense RNAs accounted for about 90% of the total identified RNAs. For each RNA category, the number of genes in AF was not observed to be higher or lower than that in control (Wilcoxon rank-sum test, P < 0.05). In addition, we also considered genes with seven specific biotypes as lncRNAs (See Material and Methods). Given a threshold of FPKM >1 in at least 25% samples (n = 3), we identified 9,233 PCGs, 2,213 lncRNAs, and 961 other ncRNA genes, which were then used for downstream analysis (**Figure 1B**).

5887 5887	8808	7530	7226	7552	8437	8494	8287	6447	8452	9269	8875	7559	group protein_coding
	993	544	513	631	972	1043	623	411	706	1013	694	520	pseudogene 1000
3095	572	269	284	308	524	624	306	212	439	509	368	253	antisense
6991	302	190	195	201	300	321	219	155	332	368	270	213	lincRNA
0931	206	128	137	141	189	208	114	81	162	188	135	93	snoRNA 100
1797	163	87	80	110	165	188	86	69	137	162	103	74	snRNA
1320	178	98	88	114	186	199	102	65	90	170	111	72	miRNA
2403	150	78	71	93	148	161	65	54	102	153	86	60	misc_RNA 10
3095	149	70	67	85	131	152	75	52	111	125	88	59	sense_intronic
5211	64	42	40	41	61	60	48	29	65	66	63	42	processed_transcript
5304	60	19	16	40	55	49	46	2	32	59	55	14	TR_J_gene 0
1704	39	45	24	32	55	30	40	21	35	17	35	34	IG_V_gene
5211	51	30	19	32	34	28	42	1	43	61	51	23	TR_V_gene group
0000	31	13	19	17	32	31	23	13	28	38	27	21	sense_overlapping
621	21	21	21	22	22	22	22	22	22	20	21	20	Mt_tRNA
0836	9	8	9	9	10	11	9	10	12	12	10	11	IG_C_gene healthy control
3705	18	3	0	7	17	15	9	7	7	13	8	7	rRNA
5284	6	5	6	6	6	6	5	3	6	7	6	5	3prime_overlapping_ncrna
4871	6	5	5	4	6	5	5	4	6	5	4	5	polymorphic_pseudogene
1047	5	5	5	5	5	5	5	4	5	5	5	5	TR_C_gene
1541	5	1	5	4	5	5	6	0	3	2	5	3	IG_J_gene
NA	2	2	2	2	2	2	2	2	2	2	2	2	Mt_rRNA
1976	2	0	1	1	0	0	1	0	2	3	1	0	TR_V_pseudogene
2012	2	0	1	1	1	1	1	0	0	1	1	0	TR_J_pseudogene
0705	1	1	1	1	1	1	1	0	0	1	0	1	IG_J_pseudogene
2824	0	0	1	0	1	1	0	0	1	0	0	0	IG_V_pseudogene
1	0	0	0	0	1	1	0	0	1	0	0	1	IG_C_pseudogene
1047 NA	0	0	0	0	0	1	0	0	0	0	0	0	TR_D_gene IG_D_gene
st P-va	lue												
	pr	otein (9	1_coc 233)										
			/	/									



Identification of Dysregulated mRNAs and IncRNAs in AF

Differential expression analysis was conducted to identify dysregulated genes in AF using the gene expression profiles. Specifically, a total of 946 genes, including 327 up- and 619 downregulated genes, were differentially expressed in AF as compared with the healthy controls (t-test, P < 0.05 and fold change >2 or $<^{1}/_{2}$, Figure 2A, Supplementary Table S1). To investigate the distinction of the dysregulated genes between AF and healthy controls, we performed hierarchical clustering analysis of the dysregulated gene expression profiles, and found that the samples with AF could be clearly distinguished from the healthy controls (Figure 2B), indicating that gene expression profiles between AF and healthy controls had marked differences. Among the dysregulated genes, the proportion of PCGs was significantly higher in the up-regulated genes than in the down-regulated genes (Figure 2C, 249/327 vs. 156/619, proportion test, *P* < 0.0001). In contrast, the proportion of lncRNAs was observed higher in the down-regulated genes than in the up-regulated genes (Figure **2C**, 315/619 vs. 63/327, proportion test, *P* < 0.0001).

Furthermore, we selected the top-five up- and down-regulated genes (**Figure 2D**), and found that the top-five up-regulated genes were all PCGs, while only one down-regulated gene, *GIMAP8*, *TNFAIP8L2*, and *RNASEL*, were involved in inflammatory response, suggesting that the dysregulation of inflammatory response may be the an important indicator for AF. On the other hand, the *PTX3* had an antiangiogenesis. In addition, lncRNAs, HOTAIRM1, RP11-262H14.1, and RP11-84A19.4, have been reported to be dysregulated in AF by previous studies (Yu et al., 2017; Qian et al., 2019). These results indicated that differential expression analysis could uncover some key genes in AF.

Gene Ontology-Based Enrichment Analysis of the Dysregulated Genes

To investigate some key biological functions involved in AF, we performed overrepresentation enrichment analysis (ORA) on the up- and down-regulated genes, respectively. We found that the up-regulated genes were highly enriched in biological functions related to response to virus, such as defense response to virus, response to virus, viral life cycle, regulation of viral process, and regulation of viral life cycle, and related to inflammatory response, such as positive regulation of I-kappaB kinase/ NF-kappaB signaling, and regulation of chemotaxis (**Figure 3A**, adjusted P < 0.05). These results indicated that chronic viral infection may lead to activated inflammatory pathways, thereby alter the electrophysiology, structure, and autonomic remodeling of the atria (Chiang et al., 2013).

Among the down-regulated GO terms, biological functions related to the response to saccharides (**Figure 3B**), such as response to lipopolysaccharide, response to glucose, response to hexose, response to monosaccharide, and response to carbohydrate were significantly enriched by the down-regulated genes. Notably, the weakened response to glucose in blood may reduce the insulin level, thereby lead to hyperglycemia, which

further demonstrate the close association between hyperglycemia and AF (Rigalleau et al., 2002).

Prediction of IncRNAs Regulating Expression of the Adjacent PCGs

It has been widely recognized that lncRNAs could regulate the expression of the adjacent PCGs by cis-acting manner (Kornienko et al., 2013). To identify these cis-acting lncRNAs, we first searched the adjacent dysregulated PCGs within one million base pairs for each dysregulated lncRNA, and found 187 lncRNA-PCG pairs. The expression levels between the lncRNA and its corresponding PCGs were highly correlated (Figure 4A). Particularly, the expression levels of about half of the lncRNA-PCG pairs were negatively correlated, indicating that the lncRNAs may suppress the expression of the adjacent PCGs. With a threshold at Pearson correlation coefficient > 0.5 or < -0.5, we identified 71 lncRNA-PCG pairs, composed of 58 PCGs and 63 lncRNAs, with potential regulatory relationship (Supplementary Table S2). Among the cis-acting lncRNAs, we found that pseudogene (46%), antisense (34%), and lincRNA (11%) were the major categories (Figure 4B).

To identify key lncRNAs involved in regulating gene expression, we selected seven lncRNAs, *AL021707.2*, *CTD-2622113.3*, *ETF1P2*, *RP11-4K3_A.5*, *RP11-95J11.1*, *ZNF137P*, and *H2AFZP1*, that regulated multiple PCGs. Notably, we found that *ETF1P2*, a pseudogene locating within 7q36, was negatively correlated with two adjacent PCGs with similar functions, *GIMAP2* and *GIMAP4* (**Figure 4C**), which participated in the regulation of T helper cell differentiation (Filen et al., 2009), indicating that the pseudogene *ETF1P2* may be the upstream regulator of T helper cell differentiation.

Functional Annotation of the Dysregulated IncRNAs by Co-Expressed PCGs

As co-expressed genes are more likely to be co-regulated, sharing similar functions, or involved in similar biological processes (Stuart et al., 2003), we predicted the function of lncRNAs by performing overrepresentation enrichment analysis on the co-expressed PCGs to identify enriched GO terms (Supplementary Table S3). We found that a large number of lncRNAs (n > 20) had the biological functions termed transcription corepressor activity, proximal promoter sequence-specific DNA binding, and RNA polymerase II proximal promoter sequence-specific DNA binding (Figure 5A). Specifically, 38 lncRNAs were characterized with transcription corepressor activity, and highly correlated with five PCGs (Pearson correlation coefficient > 0.5), SF1, MNT, NR1D1, SKI, DNAJB1, and YY1, with the same GO term (Figures 5B, C). In addition, we also found that one lncRNA, AP001053.11, may participate in inflammatory response related GO terms, such as chemokine binding, chemokine receptor activity, cytokine binding, and cytokine receptor activity (Figure 5D). Notably, three chemokine receptor, *CX3CR1*, *CCR2*, and CCR5, were highly correlated with AP001053.11 (Pearson correlation coefficient > 0.9), further suggesting a critical role of AP001053.11 in regulation of chemokine receptor activity.



FIGURE 2 | The differentially expressed genes in AF. (A) The volcano plot displays the up-regulated (red dots) and down-regulated (blue dots) genes. (B) The heatmap shows the scaled gene expression of dysregulated genes. (C) The number of PCGs, IncRNAs, and other ncRNAs in up-regulated, down-regulated, and all dysregulated genes. (D) The expression levels of the top-five up-regulated and down-regulated genes in AF and control.



respectively. The more the gene count, the larger size the circle.



DISCUSSION

LncRNAs are an emerging class of RNA species that may play a critical regulatory role in gene expression. LncRNAs can serve as diagnostic biomarkers or therapeutic targets for many diseases (Ishii et al., 2006; Chen et al., 2008; Chubb et al., 2008). However, the association between lncRNAs and AF is still not fully understood.

In this study, we used RNA sequencing data to identify and quantify the both PCGs and lncRNAs, and conducted differential expression analysis to identify dysregulated genes in AF. Specifically, a total of 946 genes, including 327 up- and 619 down-regulated genes, were differentially expressed in AF as compared with the healthy controls (t-test, P < 0.05 and fold change >2 or $<^{1}/_{2}$, **Figure 2A**, **Supplementary Table S1**). The hierarchical clustering analysis of those dysregulated gene expression profiles showed that the samples with AF could be clearly distinguished from the healthy controls (**Figure 2B**), indicating that gene expression profiles between AF and healthy controls had marked differences. Furthermore, we found that three of the top-five upregulated genes, *GIMAP8*, *TNFAIP8L2*, and *RNASEL*, were involved in inflammatory response, which was in accordance with the conclusion that the infiltration of immune cells and proteins that mediate inflammatory response in cardiac tissue and circulatory processes is associated with AF by previous studies (Yamashita et al., 2010; Harada et al., 2015). On the other hand, the *PTX3* had an antiangiogenic role, and its downregulation may lead to enhanced angiogenesis, which has been reported to be associated with AF (Berntsson et al., 2019).

To investigate some key biological functions involved in AF, we performed ORA on the dysregulated genes. The significant enrichment of these up-regulated genes in biological functions related to response to virus and inflammatory response suggested that chronic viral infection may lead to activated inflammatory pathways, thereby alter the electrophysiology, structure, and autonomic remodeling of the atria (Chiang et al., 2013). In contrast, the downregulated GO terms were related to the response to saccharides (**Figure 3B**), which gave us a hint that the weakened response to glucose in blood may reduce the insulin level, thereby lead to hyperglycemia as previous study reported (Rigalleau et al., 2002).

To identify key lncRNAs involved in AF, we predicted lncRNAregulated expression of the adjacent PCGs, and characterized





biological function of the dysregulated lncRNAs. We found that *ETF1P2*, a pseudogene locating within 7q36, was negatively correlated with two adjacent PCGs with similar functions, *GIMAP2* and *GIMAP4* (**Figure 4C**), which participated in regulation of T helper cell differentiation (Filen et al., 2009), indicating that the pseudogene *ETF1P2* may be an upstream regulator of T helper cell differentiation. Moreover, we also found that one lncRNA, *AP001053.11*, may participate in inflammatory-response-related GO terms by co-expression-based functional annotation. Notably, three chemokine receptor, *CX3CR1*, *CCR2*, and *CCR5*, were highly correlated with *AP001053.11* (Pearson correlation coefficient > 0.9), further suggesting that *AP001053.11* may be implicated in AF *via* the regulation of chemokine receptor activity.

In addition, there are also some limitations in this study. Firstly, more samples were needed to support our findings about the key lncRNAs. We will collect more samples with AF and healthy donors in the near future, which can overcome this limitation. Secondly, some experimental validation would be required for future verification of the functional lncRNAs. We hope to conduct further research with larger sample size, experimental validation and improved methodology for data analysis in the near future.

In conclusion, we have identified key PCGs and lncRNAs, which may be implicated in AF, which not only improves our understanding of the roles of lncRNAs in AF, but also provides potentially functional lncRNAs for AF researchers.

DATA AVAILABILITY STATEMENT

The datasets generated for this study can be found in the SRP093226.

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AUTHOR CONTRIBUTIONS

Conception and design: D-MW, JL, G-QC, X-WH, Y-LZ; Administrative support: JL, G-QC, Y-LZ; Provision of study materials or patients: D-MW, S-HF, Z-HZ; Collection and assembly of data: D-MW, S-HF, XW, X-RH, SW, Y-JW, Z-FZ; Data analysis and interpretation: D-MW, QS, M-QL; Manuscript writing: All authors; Final approval of manuscript: All authors.

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SUPPLEMENTARY MATERIAL

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fgene.2019.00908/full#supplementary-material

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Conflict of Interest: The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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