

Annotation of susceptibility SNPs associated with atrial fibrillation

Chengqi Xu^{1,*}, Rongfeng Zhang^{2,*}, Yunlong Xia², Liang Xiong³, Wei Yang⁴, Pengyun Wang³

¹College of Life Science and Technology, Center for Human Genome Research and Cardio-X Institute, Huazhong University of Science and Technology, Wuhan 430074, P. R. China

²Department of Cardiology, First Affiliated Hospital of Dalian Medical University, Dalian 116011, P. R. China

³Department of Clinical Laboratory, Liyuan Hospital of Tongji Medical College, Huazhong University of Science and Technology, Wuhan 430077, P. R. China

⁴Jilin Provincial Key Laboratory on Molecular and Chemical Genetic, The Second Hospital of Jilin University, Changchun 130041, P. R. China

*Equal contribution and Co-first authors

Correspondence to: Pengyun Wang, Wei Yang; email: wpy0110@mail.hust.edu.cn, wyang2002@jlu.edu.cn

Keywords: atrial fibrillation, SNP, annotation, genetics, non-coding

Received: December 16, 2019

Accepted: June 18, 2020

Published: September 9, 2020

Copyright: Xu et al. This is an open-access article distributed under the terms of the Creative Commons Attribution License (CC BY 3.0), which permits unrestricted use, distribution, and reproduction in any medium, provided the original author and source are credited.

ABSTRACT

Objective: Genome-wide association studies (GWAS) and the candidate gene based association studies have identified a panel of variants associated with atrial fibrillation (AF), however, most of the identified single nucleotide polymorphisms (SNPs) were found located within intergenic or intronic genomic regions, and whether the positive SNPs have a real biological function is unknown, and the real disease causing gene need to be studied.

Results: The current results of the genetic studies including common variants identified by GWAS (338 index SNPs) and candidate gene based association studies (40 SNPs) were summarized.

Conclusion: Our study suggests the relationship between genetic variants and possible targeted genes, and provides insight into potential genetic pathways underlying AF incidence and development. The results may provide an encyclopedia of AF susceptibility SNPs and shed light on the functional mechanisms of AF variants identified through genetic studies.

Methods: We summarized AF susceptibility SNPs identified by GWAS and candidate gene based association studies, and give a comprehensive functional annotation of all these AF susceptibility loci. by genomic annotation, microRNA binding prediction, promoter activity analysis, enhancer activity analysis, transcription factors binding activity prediction, expression quantitative trait loci (eQTL) analysis, long-range transcriptional regulatory function analysis, gene ontology and pathway enrichment analysis.

INTRODUCTION

Atrial fibrillation (AF), which is characterized by rapid and irregular beating of the atria, and known as the most common type of cardiac arrhythmia. According to the epidemiologic data, the prevalence of AF ranged from 0.7% to 1% in the general population, and up to 8% in elders greater than 80 years [1, 2]. Meanwhile, AF increases the risk of stroke, congestive heart failure,

sudden cardiac death, and increase the rate of substantial morbidity and mortality for about 2 folds [3].

AF is often associated with complications such as hypertension, valvular heart disease, coronary artery disease, heart failure, hyperthyroidism, structural heart diseases, and is also clearly heritable [4, 5]. The important role of genetic factors in the pathogenesis of

AF has shown by the identification of AF-causing mutations or rare variants in some families with lone AF, which occurs in structurally normal hearts and without known risk factors [6–8]. Meanwhile, in general AF, the non-hypothesis-driven genome-wide association studies (GWAS) and the candidate gene based association studies have identified a panel of common variants confer risk to AF [9–11]. These studies have set up a key role for the genetic background in generating for AF.

GWAS investigate associations between genomic variants and a disease or trait at the whole genome level without priori assumptions of genomic locations or potential functions of candidate genes. In this case, most of the identified single nucleotide polymorphisms (SNPs) associated with disease were found located within intergenic or intronic genomic regions, and whether the positive SNPs have a real biological function is unknown, and the real target gene need to be further studied [12, 13]. For example, SNP rs2200733 on chromosome 4p25 is the first risk variant for AF identified by GWAS and is the most robustly replicated AF locus to date. The gene that closest proximity to rs2200733 and other AF susceptibility variants in 4q25 is the *PITX2*. Studies in mice showed that *pitx2* haplo-insufficiency promotes an atrial arrhythmia [14]. However, functional evidence about the mechanisms linking these non-coding variants with *PITX2* or the incidence of AF is limited, until a recent study found that these non-coding variants in 4q25 possessing a long-range enhancer–promoter interactions and exert as a transcriptional regulatory directed function at *PITX2* [15]. Understanding the biological nature of non-coding variants associated with AF can enable us to point the real causal genes causing AF and provide insight into the mechanism of AF.

Considering one of the most important challenges of AF genetic study is to elucidate functional mechanisms that how the susceptibility loci modulate AF risk, in the current study, we summarized the results of the studies including variants identified by GWAS and candidate gene based association studies, and give a comprehensive functional annotation of all these AF susceptibility loci. The non-synonymous SNPs were first identified and classified as functional SNPs, and for SNPs in the non-coding region, we try to predict their potential functions including microRNA binding, promoter activity, enhancer activity, transcription factors binding activity, expression quantitative trait loci (eQTL), and long-range transcriptional regulatory function. Our results may provide an encyclopedia of AF susceptibility SNPs and shed light on the functional mechanisms of AF variants identified through genetic studies.

RESULTS

AF susceptibility loci

Through searching the public databases including GWAS catalog (<https://www.ebi.ac.uk/gwas/>), GWAS central (<https://www.gwascentral.org>), and literatures in Pubmed, Embase and Medline, we included 18 AF GWAS and exome-wide association study (EWAS) in our study, which published from 2007 to 2019 (Table 1). The workflow of the current study is shown in Figure 1. Participants of these studies were mainly European ancestry (15 of 18 studies), and the rest were East Asian (Korean ancestry and Japanese) (Table 1). A total of 338 SNPs (refer as index SNPs) passed the multiple corrections ($P < 5 \times 10^{-8}$ or corresponding multiple correction threshold) and showed a significant association with AF in GWAS and EWAS (Figure 2 and Supplementary Table 1). We also include 40 common SNPs which showed significant associated with AF in case control or population prospective study in candidate gene based analysis, or replication study of GWAS loci (Figure 2 and Supplementary Table 1). Totally, we included 378 AF susceptibility SNPs in our further functional annotation.

Genomic region annotation using Variant Effect Predictor (http://asia.ensembl.org/Homo_sapiens/Tools/VEP, GRCh38) showed that only a small portion of GWAS index AF SNPs was located in exon of known genes (21/338, 6.21%), 63.31% were found in introns (214/338), and 29.29% locate in intergenic regions (99/338) (Figure 3). In candidate gene based analysis, 50% identified AF related variants locate in intron, and the proportion of non-synonymous variants associated with AF (32.50%, 13/40) was higher than in GWAS index SNPs (2.936%, 10/338) (Figure 3).

Functional annotation of missense and splicing related SNPs

Through GWAS, EWAS and candidate gene based association study, a total of 36 SNPs were found located in exon of known genes. Missense variants in *SPATC1L* (rs113710653) [24], *TNFSF13* (rs11552708) [24], *SLC22A25* (rs11231397) [24], *RPL3L* (rs140185678) [16], *GCOM1* and *MYZAP* (rs147301839) [16], *UBE4B* (rs187585530) [17], *NEBL* (rs2296610) [22], *LRIG1* (rs2306272) [17], *PLEC* (rs373243633) [23], *DNAH10OS* (rs12298484) [17] were associated with risk of AF through GWAS and EWAS approach, and candidate gene based association study found missense variants in *AGTR1* [38], *AGXT2* [40], *ZFH3* [52], *MTR* [70], *KCNH2* [69], *KCNE1* [51], *NPPA* [66] confer risk to AF. Two index SNPs of GWAS, including rs140192228 in *RPL3L* [16] and rs133902 in *MYO18B* [16] were predicted may change the mRNA splicing.

Table 1. Included studies with SNPs associated with AF from 2007 to 2019.

Number	Studies	Year	Discovery population	Replication population
Genome-wide association study				
1	Nielsen JB. et al [16]	2018	60,620 European ancestry cases and 970,216 European ancestry controls	NA
2	Roselli C. et al [17]	2018	55,114 European ancestry cases and 482,295 European ancestry controls, 8,180 Japanese ancestry cases and 28,612 Japanese ancestry controls, 1,307 African American ancestry cases and 7,660 African American ancestry controls, 845 Hispanic cases and 4,177 Hispanic controls	NA
3	Nielsen JB. et al [18]	2018	6,337 European ancestry cases and 61,607 European ancestry controls	30,679 European ancestry cases and 278,895 European ancestry controls
4	Thorolfsson RB. et al [19]	2018	14,710 cases and 373,897 controls from Iceland, 14,792 cases and 393,863 controls from the UK Biobank	9,204 cases and 76,161 controls, European ancestry
5	Lee JY. et al [20]	2017	672 Korean ancestry cases and 3,700 Korean ancestry controls	200 Korean ancestry cases and 1,812 Korean ancestry controls
6	Christophersen IE. et al [21]	2017	GWAS: 18,398 individuals with atrial fibrillation and 91,536 referents, EWAS: 22,806 AF cases and 132,612 referents.	NA
7	Low SK. et al [22]	2017	8,180 Japanese ancestry cases and 28,612 Japanese ancestry controls	3,120 Japanese ancestry cases and 125,064 Japanese ancestry controls, 15,993 European ancestry cases and 113,719 European ancestry controls
8	Thorolfsson RB. et al [23]	2017	14,255 AF cases and 374,939 controls, Iceland	2,002 non-Icelandic cases and 12,324 controls
9	Yamada Y. et al [24]	2017	884 patients with atrial fibrillation and 12,282 controls, Japanese	NA
10	Lubitz SA. et al [25]	2016	1,734 individuals with and 9,423 without AF, European ancestry	NA
11	Kertai MD. et al [26]	2015	620 European ancestry cases, 257 European ancestry controls	220 cases and 84 controls
12	Sinner MF. et al [27]	2014	6,707 AF cases and 52,426 controls in Europeans, 843 AF and 3,350 controls in Japanese	6,691 AF cases and 17,144 controls in Europeans, 1,618 AF cases and 17,190 controls
13	Ellinor PT. et al [28]	2012	6,707 European ancestry cases and 52,426 European ancestry controls	5,381 European ancestry cases and 10,030 European ancestry controls
14	Ellinor PT. et al [29]	2010	1,335 European ancestry lone AF cases and 12,844 European ancestry controls	1,164 European ancestry AF cases, 3,607 European ancestry controls
15	Gudbjartsson DF. et al [30]	2009	2,385 European ancestry cases and 33,752 European ancestry controls	2,427 European ancestry cases and 3,379 European ancestry controls, 286 Han Chinese ancestry cases and 2,763 Han Chinese ancestry controls
16	Benjamin EJ. et al [31]	2009	3,413 cases and 37,105 referents, European ancestry	2,145 cases and 4,073 controls, European ancestry
17	Larson MG. et al [32]	2007	151 cases and 1,190 controls from 310 families	NA
18	Gudbjartsson DF. et al [33]	2007	550 European ancestry cases and 4,476 European ancestry controls	3,030 European ancestry cases and 14,780 European ancestry controls, 333 Han Chinese ancestry cases and 2,836 Han Chinese ancestry controls
Candidate gene based association study				
19	Cao H. et al [34]	2019	828 patients and 834 controls in Chinese population	NA
20	Xiong H. et al [35]	2019	944 AF patients and 981 non-AF controls in Chinese population	732 cases and 1,291 controls in Chinese population
21	Wang P. et al [36]	2018	1,164 AF patients and 1,460 controls	NA
21	Zaw KTT. et al [37]	2017	452 cases and 1,981 controls in Japanese	NA
22	Feng W. et al [38]	2017	300 AF cases and 300 controls	NA
23	Nakano Y. et al [39]	2016	577 cases and 1935 controls in Japanese	NA
24	Seppälä I. et al [40]	2016	1,834 individuals with AF and 7,159 unaffected individuals	NA
25	Fang Z. et al [41]	2016	1,150 AF cases and 1,150 AF-free controls in Chinese	NA
26	Wang C. et al [42]	2016	1,127 unrelated AF patients and 1,583 non-AF subjects	NA
27	Fang Z. et al [41]	2016	597 AF cases and 996 AF-free controls in Chinese	NA
28	Zhang R. et al [43]	2016	1,132 AF patients and 1,206 controls	NA
29	Roberts JD. et al [44]	2016	2,601 incident of AF in a total of 17,529 participant	NA
30	Luo Z. et al [45]	2016	889 AF patients and 1015 controls, Chinese	NA
31	Chen S. et al [46]	2015	941 cases and 562 controls, Chinese	2,113 cases and 3,381 controls

32	Liu Y. et al [47]	2015	597 AF patients and 996 non-AF controls in Chinese	NA
33	Rosenberg MA. et al [48]	2014	879 incident AF in a total 3,309 participants	NA
34	Andreasen L. et al [49]	2014	657 patients diagnosed with AF and a control group comprising 741 individuals	NA
35	Luo Z. et al [50]	2014	889 AF patients and 1,015 controls in Chinese	NA
36	Voudris K.V. et al [51]	2014	509 patients of whom 203 with AF	NA
37	Liu Y. et al [52]	2014	597 AF patients and 996 non-AF controls in Chinese	NA
37	Andreasen L. et al [49]	2014	657 AF cases and 741 controls, European ancestry	NA
38	Lin H. et al [53]	2014	948 cases and 3,330 controls, European ancestry	NA
39	Adamsson S. et al [54]	2014	520 incident AF in a total 3,900 subjects, European ancestry	2,247 cases, 2,208 controls
40	Cao H. et al [55]	2014	840 AF patients and 845 controls in Chinese	NA
41	Marott SC. et al [56]	2013	358 patients with lone AF, 299 non-lone AF, and 741 controls, European ancestry	NA
42	Jeff JM. et al [57]	2014	1,288 patients with cardiac surgery, European ancestry	NA
43	VoudrisKv. et al [51]	2014	509 patients with cardiac surgery, European ancestry	NA
44	Ilkhanoff L. et al [58]	2014	241 cases and 3,144 controls, African Americans	NA
45	Marott SC. et al [56]	2013	2,570 AF events in 69,611 participants, European ancestry	NA
46	Andreasen L. et al [59]	2013	358 patients with lone AF and a control of 751 individuals, European ancestry	NA
47	Olesen MS. et al [60]	2012	209 patients with early-onset lone AF, and a control group consisting of 534 individuals free of AF	NA
48	Schnabel RB. et al [61]	2011	European (n=18,524; 2260 AF cases in a total 18,524 individuals cohort in European ancestry), 263 AF cases in a total of 3,662 African American descent.	468 AF cases and 438 controls
49	Wirka RC. et al [62]	2011	384 early onset lone AF cases and 3,010 population control	NA
50	Li C. et al [63]	2011	650 AF patients and 1,447 non-AF controls	NA
51	Lubitz SA. et al [64]	2010	790 case and 1,177 control subjects, European ancestry	5,066 case and 30,661 referent subjects, European ancestry
52	Roberts JD. et al [65]	2010	620 AF cases and 2,446 healthy controls	NA
53	Ren X. et al [66]	2010	384 sporadic AF patients and 844 controls	NA
54	Shi L. et al [67]	2009	383 AF patients versus 851 non-AF controls	NA
55	Kääb S. et al [68]	2009	3,508 AF cases and 12,173 controls, European ancestry	NA
56	Sinner MF. et al [69]	2008	1,207 AF-cases and 2,475 controls	NA
57	Giusti B. et al [70]	2007	456 AF patients and 912 matched controls	NA

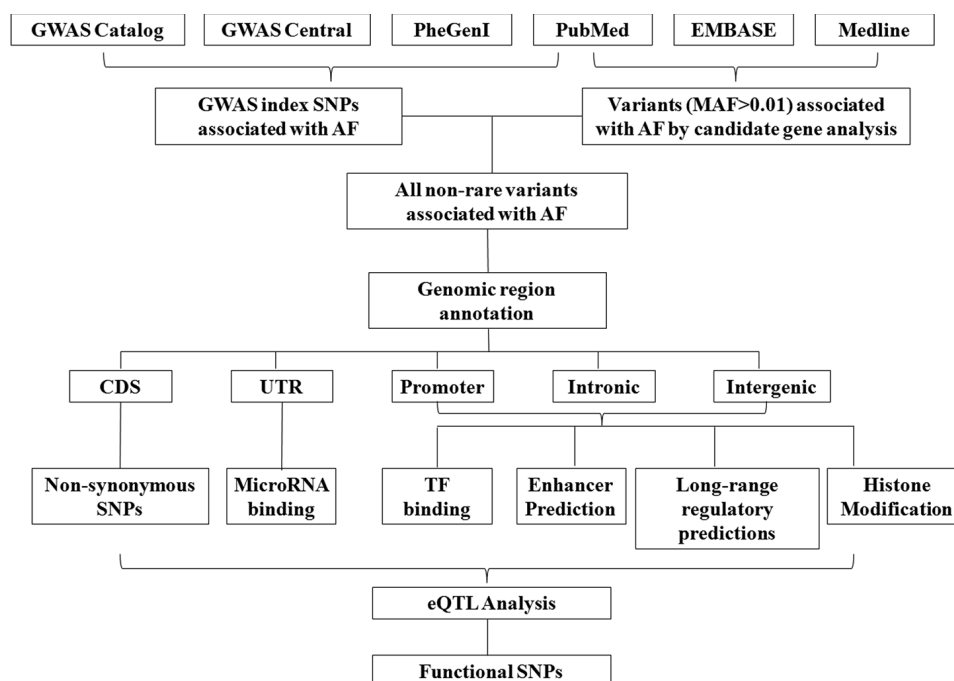


Figure 1. Workflow of the annotation of susceptibility SNPs associated with atrial fibrillation.

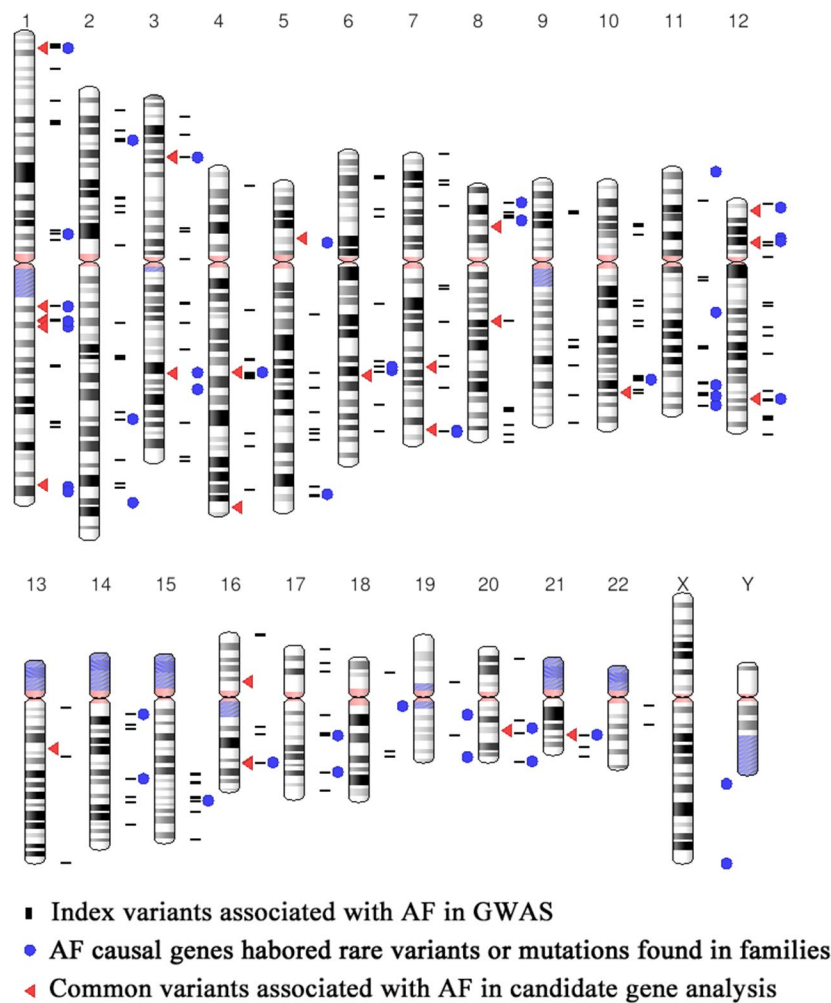


Figure 2. Distribution of the 378 AF susceptibility SNPs and AF causal genes.

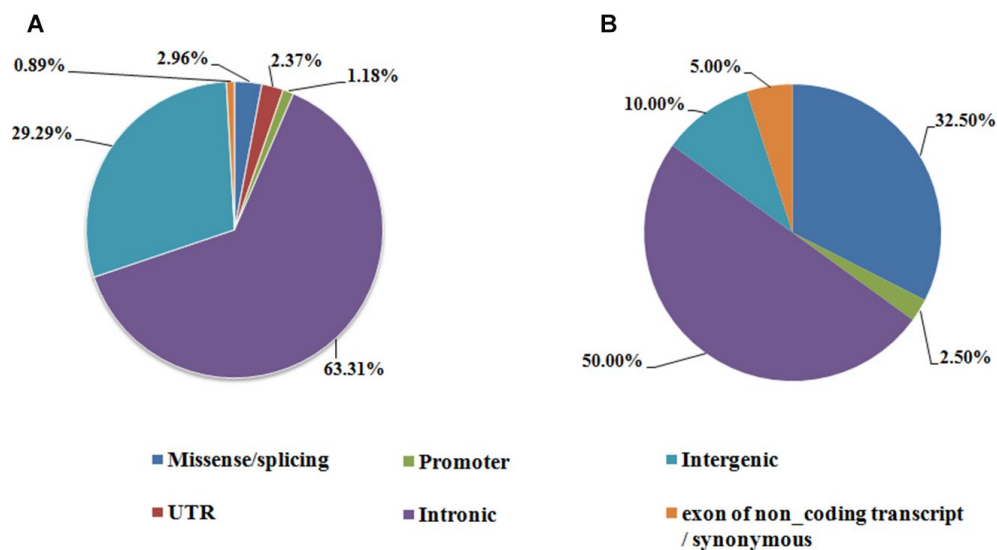


Figure 3. Genomic region annotation distribution of the AF susceptibility SNPs. (A) Index SNPs associated with AF identified in GWAS, (B) AF susceptibility SNPs identified by candidate gene based analysis.

Functional annotation of AF susceptibility SNPs in UTR

8 AF susceptibility SNPs were found in the UTR of protein-coding genes, and 4 of them were predicted to alter the microRNA binding ability predicted by MirSNP (<http://bioinfo.bjmu.edu.cn/mirsnp/search/>) and miRNASNP (<http://bioinfo.life.hust.edu.cn/>). Rs1049334 in the 3'UTR of *CAVI* was predicted change the binding with hsa-miR-125a-3p, hsa-miR-3620, hsa-miR-4299, hsa-miR-4726-3p, hsa-miR-4783-3p and hsa-miR-497-3p. Rs13385 in the 3'UTR of *HBEGF* was expected to alter the binding with hsa-miR-1207-5p and hsa-miR-4763-3p. Rs7508 in 3'UTR of *ASAH1* may change the binding of hsa-miR-134, hsa-miR-3118, hsa-miR-5190 and hsa-miR-628-5p. Rs951366 in *NUCKS1* was found in the binding region of hsa-miR-3929, hsa-miR-4419b, hsa-miR-4478, hsa-miR-4649-3p and hsa-miR-485-5p.

Functional annotation of AF susceptibility SNPs in non-coding regions

According to the data of the chromatin state and modification of histone binding, a total of 250 SNPs in non-coding regions were identified as located in enhancer regions or might affect the histone mark of promoters and enhancers (Supplementary Table 1), and further analysis found that 65 of them may change the situation of interaction with transcription factors (Supplementary Table 1). 40 transcription factors were found interact with these SNPs. After corrected by genome-wide expected binding ability, these SNPs were significantly enriched for disruption of 3 TFs including STAT6 ($P=0.02$), REST ($P=0.05$) and NFIC ($P=3.86 \times 10^{-3}$) (Figure 4).

eQTL analyses

SNPs in the non-coding region may associate with the expression levels and act as eQTL. We assessed the data from GTEx database (<https://gtexportal.org/home/>) and evaluate whether AF susceptibility SNPs affect the target gene expression levels. The results showed that 151 SNPs can affect the expression levels of a total of 328 target genes, and 81 of them associated with the expression levels of the closest gene (Supplementary Table 1). Combined with the TF binding data, 39 eQTL effect SNPs were found may alter the binding with transcription factors (Supplementary Table 1).

Long-range transcriptional regulatory function predictions

We used 3dSNP database (<http://cbportal.org/3dsnp/>) to analyze whether AF susceptibility SNPs affect distal target genes through topological interactions and

function as long-range transcriptional regulatory elements. Results indicated that a total of 211 SNPs interact with distal target genes, and 104 of them exert as an eQTL effect (Supplementary Table 1).

Gene ontology and pathway enrichment analyses of eQTL targeted genes

eQTL targeted genes of AF were mapped onto Gene ontology (GO) database using three primary categories including molecular function, protein class and biological process via PANTHER (<http://www.pantherdb.org>). The results showed that AF related genes were mainly enriched in binding, cellular process, metabolic process, protein modifying enzyme, gene-specific transcriptional regulator and membrane traffic protein (Figure 5).

All eQTL targeted genes of AF were subjected to pathway enrichment analysis in the Search Tool for the Retrieval of Interacting Genes (STRING, v11.0, <http://string-db.org>). Statistical enrichment tests were executed on gene lists within the STRING by Gene Ontology and pathway annotations. The results uncovered some signaling pathway may play roles in AF including “organelle organization”, “striated muscle cell development”, “nuclear migration”, “endomembrane system organization” and “striated muscle cell differentiation” (Table 2).

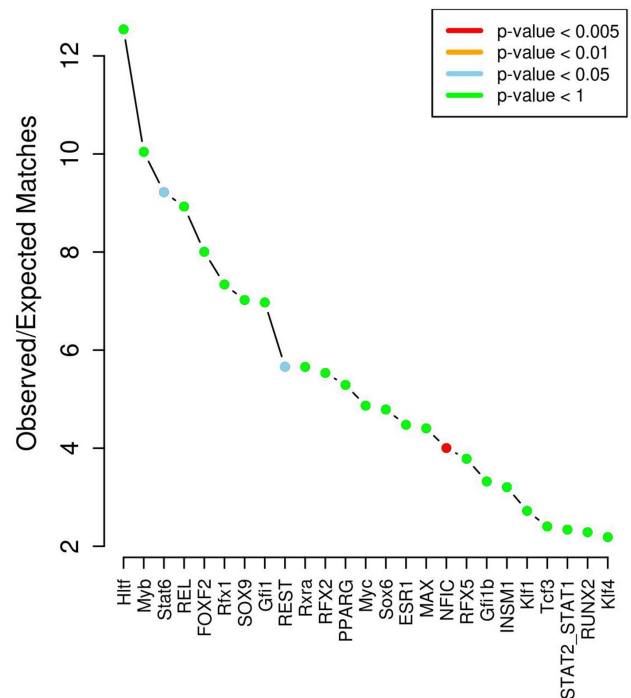


Figure 4. Transcription factor enrichment analysis results.

DISCUSSION

Population-based genetic analysis including GWAS and candidate gene based analysis has identified several SNPs associated with the risk of atrial fibrillation, here, we summarized the current results of the common

variants conferred risk to AF and totally including 378 SNPs. Considering most of the AF susceptibility SNPs were located in the non-coding genomic regions, we give a comprehensive functional annotation of all these AF susceptibility SNPs through microRNA binding prediction, promoter and enhancer activity prediction,

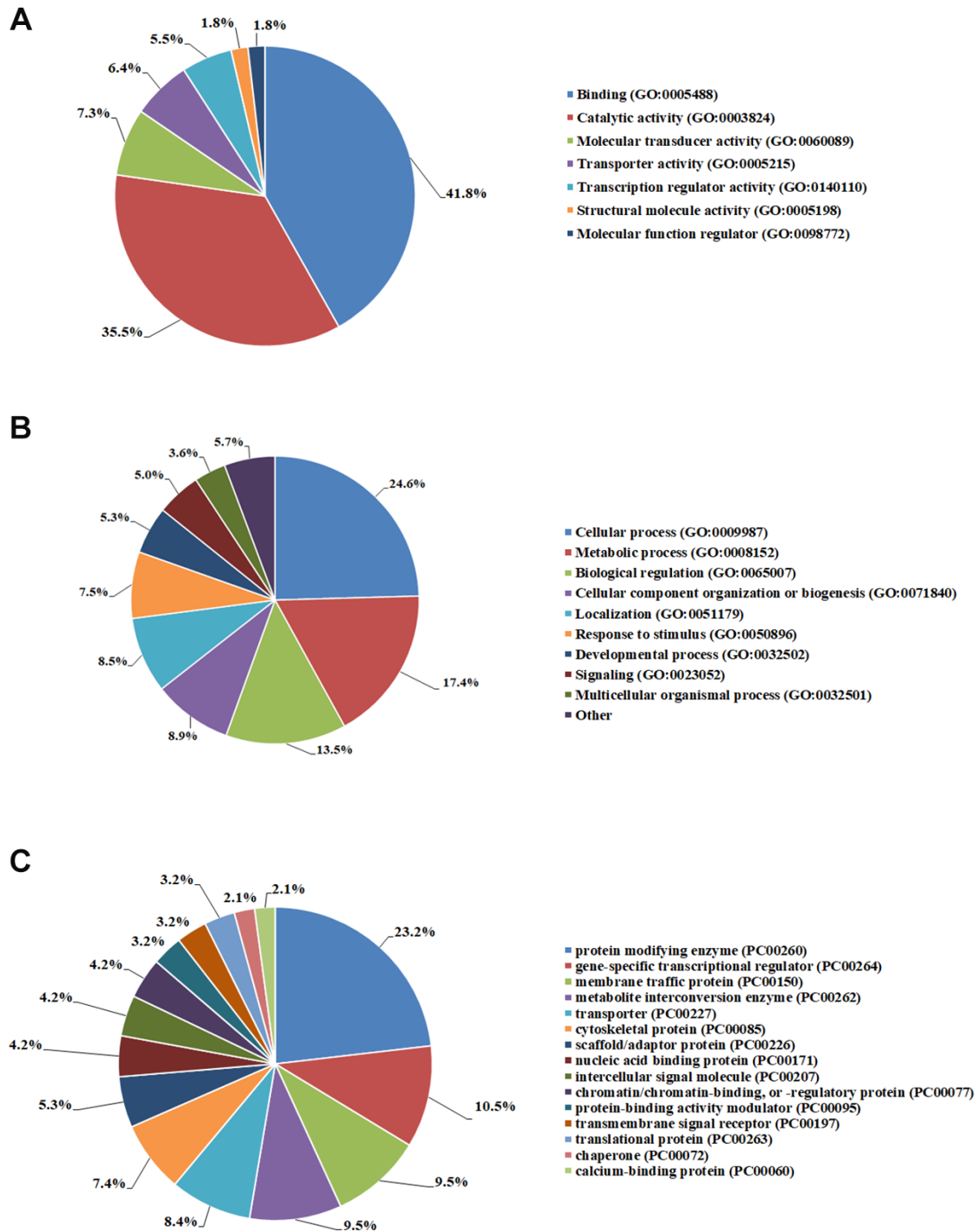


Figure 5. Gene ontology analyses of AF eQTL targeted genes. (A) Molecular function. (B) Biological process. (C) Protein class.

Table 2. Significantly enriched pathways of AF eQTL targeted genes.

#term ID	term description	observed gene count	false discovery rate	matching proteins in your network (labels)
GO:0006996	organelle organization	56	2.0x10 ⁻⁴	<i>BAZ2A, BRD8, CASQ2, CAV1, CAV2, CDC23, CEP68, CFL2, CHRAC1, CISD2, CTC1, DEK, DNM1L, GATAD1, GORAB, HDAC5, HEATR2, HIP1R, HPS6, IMMT, KANSL1, KDM1B, KIF3C, MAPT, MTHFR, MTSSI, MYH10, MYOZ1, NDUFB10, NEK6, NEURL1, NKX2-5, NR3C1, NSF, NUCKS1, PCID2, PCMI, PEX26, PFDN1, PTK2, RAB29, RAB3IP, REEP1, REEP2, REEP4, RPL3L, SCMHI, SEC24C, SYNE2, TMEM70, UBE2D3, USP36, VPS37B, WIPF1, ZNF462, ZBPB2</i>
GO:0055002	striated muscle cell development	8	0.03	<i>CASQ2, CAV2, CFL2, CHRN1, FLNC, MYH10, MYOZ1, NKX2-5</i>
GO:0007097	nuclear migration	4	0.04	<i>MYH10, PCMI, PTK2, SYNE2</i>
GO:0010256	endomembrane system organization	12	0.04	<i>CAV1, CAV2, MTSSI, MYH10, NEK6, RAB29, REEP1, REEP2, REEP4, SYNE2, VPS37B, ZBPB2</i>
GO:0051146	striated muscle cell differentiation	9	0.04	<i>BMP4, CASQ2, CAV2, CFL2, CHRN1, FLNC, MYH10, MYOZ1, NKX2-5</i>

transcription factors binding activity prediction, eQTL analysis, and long-range transcriptional regulatory function predictions.

Our functional annotation found that 151 AF susceptibility SNPs showed an eQTL effect, and 238 SNPs in non-coding regions were identified as located in enhancer regions or might affect the histone mark of promoters and enhancers. Previous studies also showed that 50-60% of the traits associated non-coding variants identified by GWAS were found located in DNase I hypersensitivity regions [71, 72], and these results also suggest that most of the SNPs identified by the GWAS as predisposing to atrial fibrillation may have biological functions and exert regulatory effects. Our results also showed that only 81 of the 151 eQTL SNPs associated with the expression levels of the closest genes, and a total of 328 target genes were identified affected by AF susceptibility SNPs. Our results identify novel genes that may be associated with the occurrence or development of AF. For example, rs35006907 located in 139bp upstream of a non-coding RNA gene LINC00964, was found associated with the expression level of *MTSSI* gene ($P=2.02 \times 10^{-18}$) in the left ventricle, which 119 kb downstream of rs35006907. Rs35006907 was predicted within an enhancer in several types of tissues including the right ventricle and right atrium, and long-range transcriptional regulatory function predictions also showed that rs35006907 and its located enhancer can interact with *MTSSI* through long-range 3D chromatin

loops. *MTSSI* can promote actin assembly at intercellular junctions and a recent functional study indicated that rs35006907 showed a cardioprotective effect [73].

Another interesting finding is about AF susceptibility loci in 10q22, which was reported as the first genetic locus for familial atrial fibrillation by Brugada R et al. in 1997, and SNPs including rs10824026 [28, 44], rs7394190 [21], rs6480708 [17] and rs60212594 [17] in 10q22 and upstream of *SYNPO2L* gene were found robustly associated with AF in several GWAS project. What is more, a missense variant in *SYNPO2L*, rs3812629 (p.Pro707Leu) was found to confer risk to AF in the Framingham population by Whole Exome Sequencing in Atrial Fibrillation [25] (Figure 6A). However, our eQTL analysis using GTEx data showed that all these GWAS positive AF SNPs including rs10824026, rs7394190, rs6480708, and rs60212594 were strong associated with the expression level of *MYOZ1* in human atrial appendage tissue with a *P* value from 1.3×10^{-28} to 1.4×10^{-45} (Figure 6B). Furthermore, the missense variant in *SYNPO2L*, rs3812629 (p.Pro707Leu), which confer risk to AF, was also found associated with *MYOZ1* expression level in human atrial appendage tissue, and the median normalized expression level of *MYOZ1* in homozygous risk allele GG carriers was -0.28 and extremely lower than in heterozygous GA carriers (0.94) (Figure 6B). *MYOZ1* encode myozenin 1, which is an intracellular binding protein involved in linking Z-disk proteins, and was known as a calcineurin-interacting protein, and help tether

calcineurin to the sarcomere of skeletal and cardiac muscle [74–76]. Mutations in *MYOZ1* were found in the patient with dilated cardiomyopathy [77–78]. These results suggested that *MYOZ1*, but not *SYNPO2L* is the causal gene of AF.

Previously genetic studies in familial or sporadic AF have identified numerous mutations or rare variants that putatively cause AF [5, 79–83], and to recently, 44 genes that putatively cause AF were mapped to pathway of ion channels/ion channels related (*ABCC9*, *HCN4*, *KCNA5*, *KCNE1*, *KCND3*, *JPH2*, *KCNE2*, *KCNE3*, *KCNE4*, *KCNE5*, *KCNH2*, *KCNJ2*, *KCNJ5*, *KCNJ8*, *KCNK3*, *KCNN3*, *KCNQ1*, *RYR2*, *SCN1B*, *SCN2B*, *SCN3B*, *SCN4B*, *SCN5A*, *SCN10A*), transcription factors (*GATA4*, *GATA5*, *GATA6*, *NKX2-5*, *NKX2-6*, *PITX2*, *SHOX2*, *SOX5*, *TBX5*, *ZFHX3*), myocardial structural components (*GJA1*, *GJA5*, *LMNA*, *MYH6*, *MYL4*, *SYNE2*), signaling, protein turnover and others (*GREM2*, *NPPA*, *SH3PXD2A*, *PLN*). Compared to the list AF susceptibility genes including the 328 eQTL target genes what we have identified and combined with the closest gene of GWAS index SNPs, only 10 genes including *HCN4*, *KCND3*, *KCNJ5*, *KCNN3*, *PITX2*, *TBX5*, *ZFHX3*, *GJA1*, *SYNE2*, *SH3PXD2A*, and *PLN* were found have both rare variants and common variants related with AF. These may result from most mutation screening were carried out in familial AF,

early-onset AF or lone AF, and AF patients in GWAS were more complex.

In conclusion, we summarized the current results of the genetic studies including common variants identified by GWAS (338 index SNPs) and candidate gene based association studies (40 SNPs), and performed a comprehensive annotation of all these AF susceptibility loci found by GWAS and candidate gene based association. We identified 4 AF susceptibility SNPs in UTRs may change the microRNA binding ability, and 250 AF susceptibility SNPs in non-coding regions were identified as located in enhancer regions or might affect the histone mark of promoters and enhancers, 65 SNPs may change the situation of interaction with transcription factors and totally 40 transcription factors were found interact with these SNPs. Our results also showed that 151 SNPs can affect the expression levels of a total of 328 target genes and 81 of them associated with the expression levels of the closest gene. Long-range transcriptional regulatory function predictions showed that 211 SNPs interact with distal target genes, and 104 of them exert as an eQTL effect. We also performed a GO and pathway enrichment of the AF eQTL genes. Taken together, our study suggested the relationship between genetic variants and possible targeted genes, and provides insight into potential genetic pathways underlying AF incidence and development.

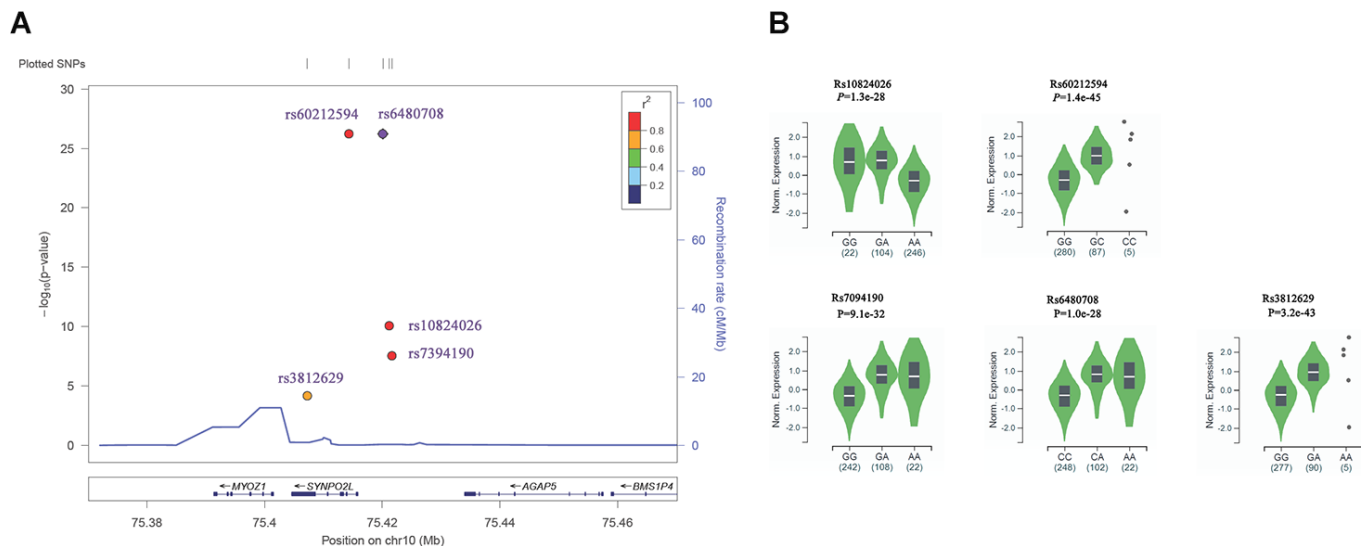


Figure 6. Association of SNPs in 10q22 with AF and eQTL analysis. (A) Regional plots for significant association with AF in 10q22. The P value was obtained from GWAS catalog database. SNPs plotted by their positions (UCSC hg19) on the corresponding chromosome against $-\log_{10}(P)$. Estimated recombination rates from 1000 genomes EUR populations were plotted in blue to reflect the local linkage disequilibrium (LD) structure on a secondary y axis. The most significant lead SNP (diamond) is denoted with the SNP identification number. Flanking SNPs (circles) are color-coded to represent the pairwise r^2 measure of LD with the lead SNP: red, $r^2 \geq 0.8$; orange, $0.6 \leq r^2 < 0.8$; green, $0.4 \leq r^2 < 0.6$; light blue, $0.2 \leq r^2 < 0.4$; blue, $r^2 < 0.2$. These plots were generated by Locuszoom (<https://statgen.sph.umich.edu/locuszoom/>). (B) eQTL analysis showed that the association between AF susceptibility SNPs in 10q22 with the expression level of *MYOZ1* in human atrial appendage tissues ($n=372$). eQTL analysis were performed using GTEx data.

MATERIALS AND METHODS

Acquisition of AF susceptibility variants and search strategy

The workflow of the current study is presented in Figure 1. First, results of the current GWAS of AF were extracted from three public databases, including GWAS catalog (<https://www.ebi.ac.uk/gwas/>), GWAS central (<https://www.gwascentral.org>) and phenotype–genotype integrator (<https://www.ncbi.nlm.nih.gov/gap/phegeni>). We also searched the literature in PubMed (<https://pubmed.ncbi.nlm.nih.gov>) to include all studies of AF GWAS. The keywords include atrial fibrillation, genome wide association or GWAS.

Besides the GWAS, several candidate gene based association studies have also identified a panel of genetic variants confer risk to AF. Results of these associated genetic variants were obtained by searching from the PubMed, EMBASE (<https://www.embase.com>) and Medline (<https://www.nlm.nih.gov/bsd/medline.html>) (Figure 1), and the searching keywords of medical subject headings (MeSH) including “atrial fibrillation” combined with “polymorphism, polymorphisms, variant, variants, single nucleotide polymorphism, single nucleotide polymorphisms, SNP, SNPs”. The results of literature searching were eligibility screened by two reviewers based on titles and abstracts. Studies published between 1 January 2007 and 1 November 2019 were included. Only case control association studies or cohort-based prospective studies were included. Functional researches, animal model studies or studies not performed in a population were excluded.

Information of AF GWAS index SNPs was extracted from the database of GWAS catalog, and the threshold of significant level for the association was set as P value lower than 5×10^{-8} . For the SNPs from candidate gene based association studies, publications were reviewed by two reviewers independently and extracted the information about the variant(s) and the details of the population. Discrepancies in data extraction were resolved by discussion or submitted to a third reviewer if required. We divided the variants analyzed in candidate gene based studies into three groups, (i) replication study of the GWAS identified susceptibility loci of AF, (ii) novel variants with minor allele frequency (MAF) $\geq 0.1\%$ (according to 1000 genome phase III global data), (iii) rare variants with a low frequency (MAF $< 0.1\%$) associated with AF by candidate gene association study or mutation screening. In our study, we excluded rare variants and mutations in (iii) from our further annotations, for the causal genes harbored mutations or rare variants of AF which were

found in families or cohort were well summarized in previously reviews [8, 84] The significant level for SNPs in candidate gene based association studies was set as satisfying the Bonferroni correction. To reduce the probability of false positives, we exclude case controls studies if the statistical power $< 70\%$. The power was extracted from publications or calculated by PS: Power and Simple Size Calculation software [85].

Genomic region annotations

All AF susceptibility SNPs including index SNPs identified by GWAS and SNPs identified by candidate gene based association studies were first annotated using Variant Effect Predictor (http://asia.ensembl.org/Homo_sapiens/Tools/VEP, GRCh38) in Ensembl to obtain their genomic region information.

Functional annotation of AF susceptibility SNPs in exon

According to the genomic region information obtained from Variant Effect Predictor, non-synonymous SNPs were directly recognized as functional variants. SNPs in untranslated region (UTR) were analyzed the microRNA binding ability using MirSNP tool (<http://bioinfo.bjmu.edu.cn/mirsnp/search/>) [86] and miRNASNP v2.0 (<http://bioinfo.life.hust.edu.cn/miRNASNP2/>) [87].

Enhancer prediction and transcription factor (TF) binding sites prediction of AF susceptibility SNPs in non-coding regions

Splicing variants identified by Variant Effect Predictor were classified as functional SNPs directly. Next, for intronic or intergenic SNPs, the chromatin states data from the Roadmap and ENCODE to analyze whether they are overlapping any enhancers in possible AF related tissues and cell types.

For the AF susceptibility SNPs in the non-coding genomic regions, including in UTR, promoter, intron, and intergenic regions, SNP2TFBS database (<http://ccg.vital-it.ch/snp2tfbs/>) was used to predict potential binding ability between SNPs and transcription factors [88].

Histone modification analysis

SNPs in promoter, intron, and intergenic regions may modify the histone binding ability, and here, using HaploReg (version 4.1) (<https://pubs.broadinstitute.org/mammals/haploreg/haploreg.php>) [89], we analyzed whether the identified non-coding AF associated SNPs overlap the major histone modifications (H3K9ac and

H3K4me3 for promoter regions, H3K27ac and H3K4me1for enhancer regions) in AF related tissues and cell types.

Long-range transcriptional regulatory function predictions

SNPs in the noncoding region may reside within or near regulatory elements controlling the expression of distal target genes through topological interactions, and using 3DSNP [90], we annotated the possible regulatory effect of identified AF associated SNPs by examining their 3D interactions with distal genes mediated by chromatin loops.

Expression quantitative trait loci analyses

Genotype-Tissue Expression (GTEx) data were used in determining whether identified AF associated SNPs affect gene expression levels. eQTL analysis were performed bases on raw RNA-Seq data (RPKM) by genes from the GTEx V6 analysis freeze (dbGaP Accession phs000424.v6.p1) and included 25 tissues.

Gene ontology (GO) and pathway enrichment analysis of eQTL targeted genes

Gene Ontology including biological process, molecular function, and protein class were annotated using PANTHER (<http://www.pantherdb.org>). KEGG pathway enrichment analysis were used in the Search Tool for the Retrieval of Interacting Genes (STRING, v11.0, <http://string-db.org>).

AUTHOR CONTRIBUTIONS

Study concept and design: Chengqi Xu, data collection: Chengqi Xu, Pengyun Wang and Rongfeng Zhang, Critical revision of the manuscript for important intellectual content: Yunlong Xia. Statistical analysis: Chengqi Xu and Rongfeng Zhang. Obtained funding: Wang PY and Wei Yang. Study supervision: Wei Yang.

CONFLICTS 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 conflicts of interest.

FUNDING

This study was supported by the National Natural Science Foundation of China Program (Grant No. 81600263, 31671302 and 81971276), National Key R&D Program of China (Grant #2018YFC1311600),

Young and Middle-aged Talents of Wuhan Municipal Health Commission (to Wang PY), Hubei health and family planning scientific research project (Grant No.WJ2019Q037).

REFERENCES

1. Bai Y, Wang YL, Shantsila A, Lip GY. The global burden of atrial fibrillation and stroke: a systematic review of the clinical epidemiology of atrial fibrillation in Asia. *Chest*. 2017; 152:810–20. <https://doi.org/10.1016/j.chest.2017.03.048> PMID:[28427968](https://pubmed.ncbi.nlm.nih.gov/28427968/)
2. Béjot Y, Ben Salem D, Osseby GV, Couvreur G, Durier J, Marie C, Cottin Y, Moreau T, Giroud M. Epidemiology of ischemic stroke from atrial fibrillation in dijon, France, from 1985 to 2006. *Neurology*. 2009; 72:346–53. <https://doi.org/10.1212/01.wnl.0000341280.31919.bd> PMID:[19171832](https://pubmed.ncbi.nlm.nih.gov/19171832/)
3. Ceonodolea AD, Bal R, Severens JL. Epidemiology and management of atrial fibrillation and stroke: review of data from four european countries. *Stroke Res Treat*. 2017; 2017:8593207. <https://doi.org/10.1155/2017/8593207> PMID:[28634569](https://pubmed.ncbi.nlm.nih.gov/28634569/)
4. Kokubo Y, Matsumoto C. Traditional cardiovascular risk factors for incident atrial fibrillation. *Circ J*. 2016; 80:2415–22. <https://doi.org/10.1253/circj.CJ-16-0919> PMID:[27795488](https://pubmed.ncbi.nlm.nih.gov/27795488/)
5. Wang P, Yang Q, Wu X, Yang Y, Shi L, Wang C, Wu G, Xia Y, Yang B, Zhang R, Xu C, Cheng X, Li S, et al. Functional dominant-negative mutation of sodium channel subunit gene SCN3B associated with atrial fibrillation in a chinese GeneID population. *Biochem Biophys Res Commun*. 2010; 398:98–104. <https://doi.org/10.1016/j.bbrc.2010.06.042> PMID:[20558140](https://pubmed.ncbi.nlm.nih.gov/20558140/)
6. Feghaly J, Zakka P, London B, MacRae CA, Refaat MM. Genetics of atrial fibrillation. *J Am Heart Assoc*. 2018; 7:e009884. <https://doi.org/10.1161/JAHA.118.009884> PMID:[30371258](https://pubmed.ncbi.nlm.nih.gov/30371258/)
7. Hayashi K, Tada H, Yamagishi M. The genetics of atrial fibrillation. *Curr Opin Cardiol*. 2017; 32:10–16. <https://doi.org/10.1097/HCO.0000000000000356> PMID:[27861186](https://pubmed.ncbi.nlm.nih.gov/27861186/)
8. Fatkin D, Santiago CF, Huttner IG, Lubitz SA, Ellinor PT. Genetics of atrial fibrillation: state of the art in 2017. *Heart Lung Circ*. 2017; 26:894–901. <https://doi.org/10.1016/j.hlc.2017.04.008> PMID:[28601532](https://pubmed.ncbi.nlm.nih.gov/28601532/)

9. Kalstø SM, Siland JE, Rienstra M, Christophersen IE. Atrial fibrillation genetics update: toward clinical implementation. *Front Cardiovasc Med*. 2019; 6:127. <https://doi.org/10.3389/fcvm.2019.00127> PMID:31552271
10. Campbell HM, Wehrens XH. Genetics of atrial fibrillation: an update. *Curr Opin Cardiol*. 2018; 33:304–10. <https://doi.org/10.1097/HCO.0000000000000505> PMID:29461262
11. Pérez-Serra A, Campuzano O, Brugada R. Update about atrial fibrillation genetics. *Curr Opin Cardiol*. 2017. [Epub ahead of print]. <https://doi.org/10.1097/HCO.0000000000000387> PMID:28169950
12. Zhu Y, Tazearslan C, Suh Y. Challenges and progress in interpretation of non-coding genetic variants associated with human disease. *Exp Biol Med* (Maywood). 2017; 242:1325–34. <https://doi.org/10.1177/1535370217713750> PMID:28581336
13. Zhang F, Lupski JR. Non-coding genetic variants in human disease. *Hum Mol Genet*. 2015; 24:R102–10. <https://doi.org/10.1093/hmg/ddv259> PMID:26152199
14. Chinchilla A, Daimi H, Lozano-Velasco E, Dominguez JN, Caballero R, Delpón E, Tamargo J, Cinca J, Hove-Madsen L, Aranega AE, Franco D. PITX2 insufficiency leads to atrial electrical and structural remodeling linked to arrhythmogenesis. *Circ Cardiovasc Genet*. 2011; 4:269–79. <https://doi.org/10.1161/CIRCGENETICS.110.958116> PMID:21511879
15. Zhang M, Hill MC, Kadow ZA, Suh JH, Tucker NR, Hall AW, Tran TT, Swinton PS, Leach JP, Margulies KB, Ellinor PT, Li N, Martin JF. Long-range Pitx2c enhancer-promoter interactions prevent predisposition to atrial fibrillation. *Proc Natl Acad Sci USA*. 2019; 116:22692–98. <https://doi.org/10.1073/pnas.1907418116> PMID:31636200
16. Nielsen JB, Thorolfsdottir RB, Fritsche LG, Zhou W, Skov MW, Graham SE, Herron TJ, McCarthy S, Schmidt EM, Sveinbjornsson G, Surakka I, Mathis MR, Yamazaki M, et al. Biobank-driven genomic discovery yields new insight into atrial fibrillation biology. *Nat Genet*. 2018; 50:1234–39. <https://doi.org/10.1038/s41588-018-0171-3> PMID:30061737
17. Roselli C, Chaffin MD, Weng LC, Aeschbacher S, Ahlberg G, Albert CM, Almgren P, Alonso A, Anderson CD, Aragam KG, Arking DE, Barnard J, Bartz TM, et al. Multi-ethnic genome-wide association study for atrial fibrillation. *Nat Genet*. 2018; 50:1225–33. <https://doi.org/10.1038/s41588-018-0133-9> PMID:29892015
18. Nielsen JB, Fritsche LG, Zhou W, Teslovich TM, Holmen OL, Gustafsson S, Gabrielsen ME, Schmidt EM, Beaumont R, Wolford BN, Lin M, Brummett CM, Preuss MH, et al. Genome-wide study of atrial fibrillation identifies seven risk loci and highlights biological pathways and regulatory elements involved in cardiac development. *Am J Hum Genet*. 2018; 102:103–15. <https://doi.org/10.1016/j.ajhg.2017.12.003> PMID:29290336
19. Thorolfsdottir RB, Sveinbjornsson G, Sulem P, Nielsen JB, Jonsson S, Halldorsson GH, Melsted P, Ivarsdottir EV, Davidsson OB, Kristjansson RP, Thorleifsson G, Helgadóttir A, Gretarsdóttir S, et al. Coding variants in RPL3L and MYZAP increase risk of atrial fibrillation. *Commun Biol*. 2018; 1:68. <https://doi.org/10.1038/s42003-018-0068-9> PMID:30271950
20. Lee JY, Kim TH, Yang PS, Lim HE, Choi EK, Shim J, Shin E, Uhm JS, Kim JS, Joung B, Oh S, Lee MH, Kim YH, Pak HN. Korean atrial fibrillation network genome-wide association study for early-onset atrial fibrillation identifies novel susceptibility loci. *Eur Heart J*. 2017; 38:2586–94. <https://doi.org/10.1093/eurheartj/ehx213> PMID:28460022
21. Christophersen IE, Rienstra M, Roselli C, Yin X, Geelhoed B, Barnard J, Lin H, Arking DE, Smith AV, Albert CM, Chaffin M, Tucker NR, Li M, et al, METASTROKE Consortium of the ISGC, Neurology Working Group of the CHARGE Consortium, and AFGen Consortium. Large-scale analyses of common and rare variants identify 12 new loci associated with atrial fibrillation. *Nat Genet*. 2017; 49:946–52. <https://doi.org/10.1038/ng.3843> PMID:28416818
22. Low SK, Takahashi A, Ebara Y, Ozaki K, Christophersen IE, Ellinor PT, Ogishima S, Yamamoto M, Satoh M, Sasaki M, Yamaji T, Iwasaki M, Tsugane S, et al, and AFGen Consortium. Identification of six new genetic loci associated with atrial fibrillation in the Japanese population. *Nat Genet*. 2017; 49:953–58. <https://doi.org/10.1038/ng.3842> PMID:28416822
23. Thorolfsdottir RB, Sveinbjornsson G, Sulem P, Helgadóttir A, Gretarsdóttir S, Benonisdóttir S, Magnusdóttir A, Davidsson OB, Rajamani S, Roden DM, Darbar D, Pedersen TR, Sabatine MS, et al. A missense variant in PLEC increases risk of atrial fibrillation. *J Am Coll Cardiol*. 2017; 70:2157–68.

- <https://doi.org/10.1016/j.jacc.2017.09.005>
PMID:[29050564](https://pubmed.ncbi.nlm.nih.gov/29050564/)
24. Yamada Y, Sakuma J, Takeuchi I, Yasukochi Y, Kato K, Oguri M, Fujimaki T, Horibe H, Muramatsu M, Sawabe M, Fujiwara Y, Taniguchi Y, Obuchi S, et al. Identification of TNFSF13, SPATC1L, SLC22A25 and SALL4 as novel susceptibility loci for atrial fibrillation by an exome-wide association study. *Mol Med Rep*. 2017; 16:5823–32.
<https://doi.org/10.3892/mmr.2017.7334>
PMID:[28849223](https://pubmed.ncbi.nlm.nih.gov/28849223/)
25. Lubitz SA, Brody JA, Bihlmeyer NA, Roselli C, Weng LC, Christophersen IE, Alonso A, Boerwinkle E, Gibbs RA, Bis JC, Cupples LA, Mohler PJ, Nickerson DA, et al, and NHLBI GO Exome Sequencing Project. Whole exome sequencing in atrial fibrillation. *PLoS Genet*. 2016; 12:e1006284.
<https://doi.org/10.1371/journal.pgen.1006284>
PMID:[27589061](https://pubmed.ncbi.nlm.nih.gov/27589061/)
26. Kertai MD, Li YJ, Ji Y, Qi W, Lombard FW, Shah SH, Kraus WE, Stafford-Smith M, Newman MF, Milano CA, Waldron N, Podgoreanu MV, Mathew JP, and Duke Perioperative Genetics and Safety Outcomes (PEGASUS) Investigative Team. Genome-wide association study of new-onset atrial fibrillation after coronary artery bypass grafting surgery. *Am Heart J*. 2015; 170:580–90.e28.
<https://doi.org/10.1016/j.ahj.2015.06.009>
PMID:[26385043](https://pubmed.ncbi.nlm.nih.gov/26385043/)
27. Sinner MF, Tucker NR, Lunetta KL, Ozaki K, Smith JG, Trompet S, Bis JC, Lin H, Chung MK, Nielsen JB, Lubitz SA, Krijthe BP, Magnani JW, et al, METASTROKE Consortium, and AFGen Consortium. Integrating genetic, transcriptional, and functional analyses to identify 5 novel genes for atrial fibrillation. *Circulation*. 2014; 130:1225–35.
<https://doi.org/10.1161/CIRCULATIONAHA.114.009892>
PMID:[25124494](https://pubmed.ncbi.nlm.nih.gov/25124494/)
28. Ellinor PT, Lunetta KL, Albert CM, Glazer NL, Ritchie MD, Smith AV, Arking DE, Müller-Nurasyid M, Krijthe BP, Lubitz SA, Bis JC, Chung MK, Dörr M, et al. Meta-analysis identifies six new susceptibility loci for atrial fibrillation. *Nat Genet*. 2012; 44:670–75.
<https://doi.org/10.1038/ng.2261>
PMID:[22544366](https://pubmed.ncbi.nlm.nih.gov/22544366/)
29. Ellinor PT, Lunetta KL, Glazer NL, Pfeufer A, Alonso A, Chung MK, Sinner MF, de Bakker PI, Mueller M, Lubitz SA, Fox E, Darbar D, Smith NL, et al. Common variants in KCNN3 are associated with lone atrial fibrillation. *Nat Genet*. 2010; 42:240–44.
<https://doi.org/10.1038/ng.537> PMID:[20173747](https://pubmed.ncbi.nlm.nih.gov/20173747/)
30. Gudbjartsson DF, Holm H, Gretarsdottir S, Thorleifsson G, Walters GB, Thorgeirsson G, Gulcher J, Mathiesen EB, Njølstad I, Nyrnes A, Wilsgaard T, Hald EM, Hveem K, et al. A sequence variant in ZFX3 on 16q22 associates with atrial fibrillation and ischemic stroke. *Nat Genet*. 2009; 41:876–78.
<https://doi.org/10.1038/ng.417>
PMID:[19597491](https://pubmed.ncbi.nlm.nih.gov/19597491/)
31. Benjamin EJ, Rice KM, Arking DE, Pfeufer A, van Noord C, Smith AV, Schnabel RB, Bis JC, Boerwinkle E, Sinner MF, Dehghan A, Lubitz SA, D'Agostino RB Sr, et al. Variants in ZFX3 are associated with atrial fibrillation in individuals of european ancestry. *Nat Genet*. 2009; 41:879–81.
<https://doi.org/10.1038/ng.416>
PMID:[19597492](https://pubmed.ncbi.nlm.nih.gov/19597492/)
32. Larson MG, Atwood LD, Benjamin EJ, Cupples LA, D'Agostino RB Sr, Fox CS, Govindaraju DR, Guo CY, Heard-Costa NL, Hwang SJ, Murabito JM, Newton-Cheh C, O'Donnell CJ, et al. Framingham heart study 100K project: genome-wide associations for cardiovascular disease outcomes. *BMC Med Genet*. 2007 (Suppl 1); 8:S5.
<https://doi.org/10.1186/1471-2350-8-S1-S5>
PMID:[17903304](https://pubmed.ncbi.nlm.nih.gov/17903304/)
33. Gudbjartsson DF, Arnar DO, Helgadóttir A, Gretarsdóttir S, Holm H, Sigurdsson A, Jonasdóttir A, Baker A, Thorleifsson G, Kristjansson K, Pálsson A, Blondal T, Sulem P, et al. Variants conferring risk of atrial fibrillation on chromosome 4q25. *Nature*. 2007; 448:353–57.
<https://doi.org/10.1038/nature06007>
PMID:[17603472](https://pubmed.ncbi.nlm.nih.gov/17603472/)
34. Cao H, Xu W, Chen X, Zhou Q, Lan R, Chen Y, Wang D. Functional promoter -1816C>G variant of RANKL predicts risk and prognosis of lone atrial fibrillation. *Heart Vessels*. 2019; 34:151–58.
<https://doi.org/10.1007/s00380-018-1222-5>
PMID:[30043156](https://pubmed.ncbi.nlm.nih.gov/30043156/)
35. Xiong H, Yang Q, Zhang X, Wang P, Chen F, Liu Y, Wang P, Zhao Y, Li S, Huang Y, Chen S, Wang X, Zhang H, et al. Significant association of rare variant p.Gly8Ser in cardiac sodium channel β 4-subunit SCN4B with atrial fibrillation. *Ann Hum Genet*. 2019; 83:239–48.
<https://doi.org/10.1111/ahg.12305>
PMID:[30821358](https://pubmed.ncbi.nlm.nih.gov/30821358/)
36. Wang P, Qin W, Wang P, Huang Y, Liu Y, Zhang R, Li S, Yang Q, Wang X, Chen F, Liu J, Yang B, Cheng X, et al. Genomic variants in NEURL, GJA1 and CUX2 significantly increase genetic susceptibility to atrial fibrillation. *Sci Rep*. 2018; 8:3297.
<https://doi.org/10.1038/s41598-018-21611-7>
PMID:[29459676](https://pubmed.ncbi.nlm.nih.gov/29459676/)
37. Zaw KT, Sato N, Ikeda S, Thu KS, Mieno MN, Arai T, Mori S, Furukawa T, Sasano T, Sawabe M, Tanaka M,

- Muramatsu M. Association of ZFH3 gene variation with atrial fibrillation, cerebral infarction, and lung thromboembolism: an autopsy study. *J Cardiol*. 2017; 70:180–84.
<https://doi.org/10.1016/j.ijcc.2016.11.005>
PMID:28007413
38. Feng W, Sun L, Qu XF. Association of AGTR1 and ACE2 gene polymorphisms with structural atrial fibrillation in a chinese han population. *Pharmazie*. 2017; 72:17–21.
<https://doi.org/10.1691/ph.2017.6752>
PMID:29441892
39. Nakano Y, Ochi H, Onohara Y, Sairaku A, Tokuyama T, Matsumura H, Tomomori S, Amioka M, Hironomobe N, Motoda C, Oda N, Chayama K, Chen CH, et al. Genetic variations of aldehyde dehydrogenase 2 and alcohol dehydrogenase 1B are associated with the etiology of atrial fibrillation in Japanese. *J Biomed Sci*. 2016; 23:89.
<https://doi.org/10.1186/s12929-016-0304-x>
PMID:27927211
40. Seppälä I, Kleber ME, Bevan S, Lyytikäinen LP, Oksala N, Hernesniemi JA, Mäkelä KM, Rothwell PM, Sudlow C, Dichgans M, Mononen N, Vlachopoulou E, Sinisalo J, et al. Associations of functional alanine-glyoxylate aminotransferase 2 gene variants with atrial fibrillation and ischemic stroke. *Sci Rep*. 2016; 6:23207.
<https://doi.org/10.1038/srep23207>
PMID:26984639
41. Fang Z, Jiang Y, Wang Y, Lin Y, Liu Y, Zhao L, Xu Y, Toorabally MB, He S, Zhang F. The rs6771157 C/G polymorphism in SCN10A is associated with the risk of atrial fibrillation in a chinese han population. *Sci Rep*. 2016; 6:35212.
<https://doi.org/10.1038/srep35212>
PMID:27725708
42. Wang C, Wu M, Qian J, Li B, Tu X, Xu C, Li S, Chen S, Zhao Y, Huang Y, Shi L, Cheng X, Liao Y, et al. Identification of rare variants in TNNI3 with atrial fibrillation in a chinese GenE D population. *Mol Genet Genomics*. 2016; 291:79–92.
<https://doi.org/10.1007/s00438-015-1090-y>
PMID:26169204
43. Zhang R, Tian X, Gao L, Li H, Yin X, Dong Y, Yang Y, Xia Y. Common variants in the TBX5 gene associated with atrial fibrillation in a chinese han population. *PLoS One*. 2016; 11:e0160467.
<https://doi.org/10.1371/journal.pone.0160467>
PMID:27479212
44. Roberts JD, Hu D, Heckbert SR, Alonso A, Dewland TA, Vittinghoff E, Liu Y, Psaty BM, Olgin JE, Magnani JW, Huntsman S, Burchard EG, Arking DE, et al. Genetic investigation into the differential risk of atrial fibrillation among black and white individuals. *JAMA Cardiol*. 2016; 1:442–50.
<https://doi.org/10.1001/jamacardio.2016.1185>
PMID:27438321
45. Luo Z, Yan C, Yu P, Bao W, Shen X, Zheng W, Lin X, Wang Z, Chen H, Chen F, Liu D, Huang M. CASP3 genetic variants and susceptibility to atrial fibrillation in chinese han population. *Int J Cardiol*. 2015; 183:1–5.
<https://doi.org/10.1016/j.ijcard.2015.01.048>
PMID:25662045
46. Chen S, Wang C, Wang X, Xu C, Wu M, Wang P, Tu X, Wang QK. Significant association between CAV1 variant rs3807989 on 7p31 and atrial fibrillation in a chinese han population. *J Am Heart Assoc*. 2015; 4:e001980.
<https://doi.org/10.1161/JAHA.115.001980>
PMID:25953654
47. Liu Y, Ni B, Lin Y, Chen XG, Chen M, Hu Z, Zhang F. The rs3807989 g/a polymorphism in CAV1 is associated with the risk of atrial fibrillation in chinese han populations. *Pacing Clin Electrophysiol*. 2015; 38:164–70.
<https://doi.org/10.1111/pace.12494>
PMID:25196315
48. Rosenberg MA, Kaplan RC, Siscovick DS, Psaty BM, Heckbert SR, Newton-Cheh C, Mukamal KJ. Genetic variants related to height and risk of atrial fibrillation: the cardiovascular health study. *Am J Epidemiol*. 2014; 180:215–22.
<https://doi.org/10.1093/aje/kwu126> PMID:24944287
49. Andreassen L, Nielsen JB, Darkner S, Christophersen IE, Jabbari J, Refsgaard L, Thiis JJ, Sajadieh A, Tveit A, Haunsø S, Svendsen JH, Schmitt N, Olesen MS. Brugada syndrome risk loci seem protective against atrial fibrillation. *Eur J Hum Genet*. 2014; 22:1357–61.
<https://doi.org/10.1038/ejhg.2014.46>
PMID:24667784
50. Luo Z, Yan C, Zhang W, Shen X, Zheng W, Chen F, Cao X, Yang Y, Lin X, Wang Z, Huang M. Association between SNP rs13376333 and rs1131820 in the KCNN3 gene and atrial fibrillation in the chinese han population. *Clin Chem Lab Med*. 2014; 52:1867–73.
<https://doi.org/10.1515/cclm-2014-0491>
PMID:24978901
51. Voudris KV, Apostolakis S, Karyofyllis P, Doukas K, Zaravinos A, Androutsopoulos VP, Michalis A, Voudris V, Spandidos DA. Genetic diversity of the KCNE1 gene and susceptibility to postoperative atrial fibrillation. *Am Heart J*. 2014; 167:274–80.e1.
<https://doi.org/10.1016/j.ahj.2013.09.020>
PMID:24439990
52. Liu Y, Ni B, Lin Y, Chen XG, Fang Z, Zhao L, Hu Z, Zhang F. Genetic polymorphisms in ZFH3 are associated with atrial fibrillation in a chinese han population. *PLoS One*. 2014; 9:e101318.

- <https://doi.org/10.1371/journal.pone.0101318>
PMID:24983873
53. Lin H, Sinner MF, Brody JA, Arking DE, Lunetta KL, Rienstra M, Lubitz SA, Magnani JW, Sotoodehnia N, McKnight B, McManus DD, Boerwinkle E, Psaty BM, et al, and CHARGE Atrial Fibrillation Working Group. Targeted sequencing in candidate genes for atrial fibrillation: the cohorts for heart and aging research in genomic epidemiology (CHARGE) targeted sequencing study. *Heart Rhythm*. 2014; 11:452–57.
<https://doi.org/10.1016/j.hrthm.2013.11.012>
PMID:24239840
54. Adamsson Eryd S, Sjögren M, Smith JG, Nilsson PM, Melander O, Hedblad B, Engström G. Ceruloplasmin and atrial fibrillation: evidence of causality from a population-based mendelian randomization study. *J Intern Med*. 2014; 275:164–71.
<https://doi.org/10.1111/joim.12144>
PMID:24118451
55. Cao H, Zhou Q, Lan R, Røe OD, Chen X, Chen Y, Wang D. A functional polymorphism C-509T in TGFβ-1 promoter contributes to susceptibility and prognosis of lone atrial fibrillation in chinese population. *PLoS One*. 2014; 9:e112912.
<https://doi.org/10.1371/journal.pone.0112912>
PMID:25402477
56. Marott SC, Nordestgaard BG, Jensen GB, Tybjaerg-Hansen A, Benn M. AT1 mutations and risk of atrial fibrillation based on genotypes from 71,000 individuals from the general population. *Br J Clin Pharmacol*. 2013; 76:114–24.
<https://doi.org/10.1111/bcp.12050> PMID:23210602
57. Jeff JM, Donahue BS, Brown-Gentry K, Roden DM, Crawford DC, Stein CM, Kurnik D. Genetic variation in the β1-adrenergic receptor is associated with the risk of atrial fibrillation after cardiac surgery. *Am Heart J*. 2014; 167:101–08.e1.
<https://doi.org/10.1016/j.ahj.2013.09.016>
PMID:24332148
58. Ilkhanoff L, Arking DE, Lemaitre RN, Alonso A, Chen LY, Durda P, Hesselson SE, Kerr KF, Magnani JW, Marcus GM, Schnabel RB, Smith JG, Soliman EZ, et al, Candidate-Gene Association Resource (CARE) Consortium and the Cardiac Arrest Blood Study (CABS) Investigators. A common SCN5A variant is associated with PR interval and atrial fibrillation among african americans. *J Cardiovasc Electrophysiol*. 2014; 25:1150–57.
<https://doi.org/10.1111/jce.12483>
PMID:25065297
59. Andreasen L, Nielsen JB, Christophersen IE, Holst AG, Sajadieh A, Tveit A, Haunsø S, Svendsen JH, Schmitt N, Olesen MS. Genetic modifier of the QTc interval associated with early-onset atrial fibrillation. *Can J Cardiol*. 2013; 29:1234–40.
<https://doi.org/10.1016/j.cjca.2013.06.009>
PMID:24074973
60. Olesen MS, Holst AG, Jabbari J, Nielsen JB, Christophersen IE, Sajadieh A, Haunsø S, Svendsen JH. Genetic loci on chromosomes 4q25, 7p31, and 12p12 are associated with onset of lone atrial fibrillation before the age of 40 years. *Can J Cardiol*. 2012; 28:191–95.
<https://doi.org/10.1016/j.cjca.2011.11.016>
PMID:22336519
61. Schnabel RB, Kerr KF, Lubitz SA, Alkylbekova EL, Marcus GM, Sinner MF, Magnani JW, Wolf PA, Deo R, Lloyd-Jones DM, Lunetta KL, Mehra R, Levy D, et al, and Candidate Gene Association Resource (CARE) Atrial Fibrillation/Electrocardiography Working Group. Large-scale candidate gene analysis in whites and african americans identifies IL6R polymorphism in relation to atrial fibrillation: the national heart, lung, and blood institute's candidate gene association resource (CARE) project. *Circ Cardiovasc Genet*. 2011; 4:557–64.
<https://doi.org/10.1161/CIRCGENETICS.110.959197>
PMID:21846873
62. Wirka RC, Gore S, Van Wagoner DR, Arking DE, Lubitz SA, Lunetta KL, Benjamin EJ, Alonso A, Ellinor PT, Barnard J, Chung MK, Smith JD. A common connexin-40 gene promoter variant affects connexin-40 expression in human atria and is associated with atrial fibrillation. *Circ Arrhythm Electrophysiol*. 2011; 4:87–93.
<https://doi.org/10.1161/CIRCEP.110.959726>
PMID:21076161
63. Li C, Wang F, Yang Y, Fu F, Xu C, Shi L, Li S, Xia Y, Wu G, Cheng X, Liu H, Wang C, Wang P, et al. Significant association of SNP rs2106261 in the ZFX3 gene with atrial fibrillation in a chinese han GenID population. *Hum Genet*. 2011; 129:239–46.
<https://doi.org/10.1007/s00439-010-0912-6>
PMID:21107608
64. Lubitz SA, Sinner MF, Lunetta KL, Makino S, Pfeufer A, Rahman R, Veltman CE, Barnard J, Bis JC, Danik SP, Sonni A, Shea MA, Del Monte F, et al. Independent susceptibility markers for atrial fibrillation on chromosome 4q25. *Circulation*. 2010; 122:976–84.
<https://doi.org/10.1161/CIRCULATIONAHA.109.886440>
PMID:20733104
65. Roberts JD, Davies RW, Lubitz SA, Thibodeau IL, Nery PB, Birnie DH, Benjamin EJ, Lemery R, Ellinor PT, Gollob MH. Evaluation of non-synonymous NPPA single nucleotide polymorphisms in atrial fibrillation. *Europace*. 2010; 12:1078–83.

- <https://doi.org/10.1093/europace/euq161>
PMID:[20543198](https://pubmed.ncbi.nlm.nih.gov/20543198/)
66. Ren X, Xu C, Zhan C, Yang Y, Shi L, Wang F, Wang C, Xia Y, Yang B, Wu G, Wang P, Li X, Wang D, et al. Identification of NPPA variants associated with atrial fibrillation in a chinese GeneID population. *Clin Chim Acta*. 2010; 411:481–85.
<https://doi.org/10.1016/j.cca.2009.12.019>
PMID:[20064500](https://pubmed.ncbi.nlm.nih.gov/20064500/)
67. Shi L, Li C, Wang C, Xia Y, Wu G, Wang F, Xu C, Wang P, Li X, Wang D, Xiong X, Bai Y, Liu M, et al. Assessment of association of rs2200733 on chromosome 4q25 with atrial fibrillation and ischemic stroke in a chinese han population. *Hum Genet*. 2009; 126:843–49.
<https://doi.org/10.1007/s00439-009-0737-3>
PMID:[19707791](https://pubmed.ncbi.nlm.nih.gov/19707791/)
68. Kääh S, Darbar D, van Noord C, Dupuis J, Pfeufer A, Newton-Cheh C, Schnabel R, Makino S, Sinner MF, Kannankeril PJ, Beckmann BM, Choudry S, Donahue BS, et al. Large scale replication and meta-analysis of variants on chromosome 4q25 associated with atrial fibrillation. *Eur Heart J*. 2009; 30:813–19.
<https://doi.org/10.1093/eurheartj/ehn578>
PMID:[19141561](https://pubmed.ncbi.nlm.nih.gov/19141561/)
69. Sinner MF, Pfeufer A, Akyol M, Beckmann BM, Hinterseer M, Wacker A, Perz S, Sauter W, Illig T, Näbauer M, Schmitt C, Wichmann HE, Schömig A, et al. The non-synonymous coding IKr-channel variant KCNH2-K897T is associated with atrial fibrillation: results from a systematic candidate gene-based analysis of KCNH2 (HERG). *Eur Heart J*. 2008; 29:907–14.
<https://doi.org/10.1093/eurheartj/ehm619>
PMID:[18222980](https://pubmed.ncbi.nlm.nih.gov/18222980/)
70. Giusti B, Gori AM, Marcucci R, Sestini I, Saracini C, Sticchi E, Gensini F, Fatini C, Abbate R, Gensini GF. Role of C677T and A1298C MTHFR, A2756G MTR and -786 C/T eNOS gene polymorphisms in atrial fibrillation susceptibility. *PLoS One*. 2007; 2:e495.
<https://doi.org/10.1371/journal.pone.0000495>
PMID:[17551576](https://pubmed.ncbi.nlm.nih.gov/17551576/)
71. Maurano MT, Humbert R, Rynes E, Thurman RE, Haugen E, Wang H, Reynolds AP, Sandstrom R, Qu H, Brody J, Shafer A, Neri F, Lee K, et al. Systematic localization of common disease-associated variation in regulatory DNA. *Science*. 2012; 337:1190–95.
<https://doi.org/10.1126/science.1222794>
PMID:[22955828](https://pubmed.ncbi.nlm.nih.gov/22955828/)
72. Dong SS, Zhang YJ, Chen YX, Yao S, Hao RH, Rong Y, Niu HM, Chen JB, Guo Y, Yang TL. Comprehensive review and annotation of susceptibility SNPs associated with obesity-related traits. *Obes Rev*. 2018; 19:917–30.
<https://doi.org/10.1111/obr.12677>
PMID:[29527783](https://pubmed.ncbi.nlm.nih.gov/29527783/)
73. Morley MP, Wang X, Hu R, Brandimarto J, Tucker NR, Felix JF, Smith NL, van der Harst P, Ellinor PT, Margulies KB, Musunuru K, Cappola TP. Cardioprotective effects of MTSS1 enhancer variants. *Circulation*. 2019; 139:2073–76.
<https://doi.org/10.1161/CIRCULATIONAHA.118.037939>
PMID:[31070942](https://pubmed.ncbi.nlm.nih.gov/31070942/)
74. Gontier Y, Taivainen A, Fontao L, Sonnenberg A, van der Flier A, Carpen O, Faulkner G, Borradori L. The z-disc proteins myotilin and FATZ-1 interact with each other and are connected to the sarcolemma via muscle-specific filamins. *J Cell Sci*. 2005; 118:3739–49.
<https://doi.org/10.1242/jcs.02484>
PMID:[16076904](https://pubmed.ncbi.nlm.nih.gov/16076904/)
75. Takada F, Vander Woude DL, Tong HQ, Thompson TG, Watkins SC, Kunkel LM, Beggs AH. Myozenin: an alpha-actinin- and gamma-filamin-binding protein of skeletal muscle Z lines. *Proc Natl Acad Sci USA*. 2001; 98:1595–600.
<https://doi.org/10.1073/pnas.041609698>
PMID:[11171996](https://pubmed.ncbi.nlm.nih.gov/11171996/)
76. Roberts MD, Romero MA, Mobley CB, Mumford PW, Roberson PA, Haun CT, Vann CG, Osburn SC, Holmes HH, Greer RA, Lockwood CM, Parry HA, Kavazis AN. Skeletal muscle mitochondrial volume and myozenin-1 protein differences exist between high versus low anabolic responders to resistance training. *PeerJ*. 2018; 6:e5338.
<https://doi.org/10.7717/peerj.5338>
PMID:[30065891](https://pubmed.ncbi.nlm.nih.gov/30065891/)
77. Posch MG, Perrot A, Dietz R, Ozcelik C, Pankuweit S, Ruppert V, Richter A, Maisch B. Mutations in MYOZ1 as well as MYOZ2 encoding the calsarcins are not associated with idiopathic and familial dilated cardiomyopathy. *Mol Genet Metab*. 2007; 91:207–08.
<https://doi.org/10.1016/j.ymgme.2007.02.014>
PMID:[17434779](https://pubmed.ncbi.nlm.nih.gov/17434779/)
78. Arola AM, Sanchez X, Murphy RT, Hasle E, Li H, Elliott PM, McKenna WJ, Towbin JA, Bowles NE. Mutations in PDLIM3 and MYOZ1 encoding myocyte Z line proteins are infrequently found in idiopathic dilated cardiomyopathy. *Mol Genet Metab*. 2007; 90:435–40.
<https://doi.org/10.1016/j.ymgme.2006.12.008>
PMID:[17254821](https://pubmed.ncbi.nlm.nih.gov/17254821/)
79. Chen YH, Xu SJ, Bendahhou S, Wang XL, Wang Y, Xu WY, Jin HW, Sun H, Su XY, Zhuang QN, Yang YQ, Li YB, Liu Y, et al. KCNQ1 gain-of-function mutation in familial atrial fibrillation. *Science*. 2003; 299:251–54.
<https://doi.org/10.1126/science.1077771>
PMID:[12522251](https://pubmed.ncbi.nlm.nih.gov/12522251/)
80. Yang Y, Xia M, Jin Q, Bendahhou S, Shi J, Chen Y, Liang B, Lin J, Liu Y, Liu B, Zhou Q, Zhang D, Wang R, et al. Identification of a KCNE2 gain-of-function mutation in

- patients with familial atrial fibrillation. *Am J Hum Genet.* 2004; 75:899–905.
<https://doi.org/10.1086/425342>
PMID:[15368194](https://pubmed.ncbi.nlm.nih.gov/15368194/)
81. Ravn LS, Aizawa Y, Pollevick GD, Hofman-Bang J, Cordeiro JM, Dixen U, Jensen G, Wu Y, Burashnikov E, Haunso S, Guerchicoff A, Hu D, Svendsen JH, et al. Gain of function in IKs secondary to a mutation in KCNE5 associated with atrial fibrillation. *Heart Rhythm.* 2008; 5:427–35.
<https://doi.org/10.1016/j.hrthm.2007.12.019>
PMID:[18313602](https://pubmed.ncbi.nlm.nih.gov/18313602/)
82. Tsai CT, Lai LP, Hwang JJ, Lin JL, Chiang FT. Molecular genetics of atrial fibrillation. *J Am Coll Cardiol.* 2008; 52:241–50.
<https://doi.org/10.1016/j.jacc.2008.02.072>
PMID:[18634977](https://pubmed.ncbi.nlm.nih.gov/18634977/)
83. Bartos DC, Duchatelet S, Burgess DE, Klug D, Denjoy I, Peat R, Lupoglazoff JM, Fressart V, Berthet M, Ackerman MJ, January CT, Guicheney P, Delisle BP. R231C mutation in KCNQ1 causes long QT syndrome type 1 and familial atrial fibrillation. *Heart Rhythm.* 2011; 8:48–55.
<https://doi.org/10.1016/j.hrthm.2010.09.010>
PMID:[20850564](https://pubmed.ncbi.nlm.nih.gov/20850564/)
84. Xiao J, Liang D, Chen YH. The genetics of atrial fibrillation: from the bench to the bedside. *Annu Rev Genomics Hum Genet.* 2011; 12:73–96.
<https://doi.org/10.1146/annurev-genom-082410-101515>
PMID:[21682648](https://pubmed.ncbi.nlm.nih.gov/21682648/)
85. Dupont WD, Plummer WD Jr. Power and sample size calculations for studies involving linear regression. *Control Clin Trials.* 1998; 19:589–601.
[https://doi.org/10.1016/s0197-2456\(98\)00037-3](https://doi.org/10.1016/s0197-2456(98)00037-3)
PMID:[9875838](https://pubmed.ncbi.nlm.nih.gov/9875838/)
86. Liu C, Zhang F, Li T, Lu M, Wang L, Yue W, Zhang D. MirSNP, a database of polymorphisms altering miRNA target sites, identifies miRNA-related SNPs in GWAS SNPs and eQTLs. *BMC Genomics.* 2012; 13:661.
<https://doi.org/10.1186/1471-2164-13-661>
PMID:[23173617](https://pubmed.ncbi.nlm.nih.gov/23173617/)
87. Gong J, Liu C, Liu W, Wu Y, Ma Z, Chen H, Guo AY. An update of miRNASNP database for better SNP selection by GWAS data, miRNA expression and online tools. *Database (Oxford).* 2015; 2015:bav029.
<https://doi.org/10.1093/database/bav029>
PMID:[25877638](https://pubmed.ncbi.nlm.nih.gov/25877638/)
88. Kumar S, Ambrosini G, Bucher P. SNP2TFBS - a database of regulatory SNPs affecting predicted transcription factor binding site affinity. *Nucleic Acids Res.* 2017; 45:D139–44.
<https://doi.org/10.1093/nar/gkw1064>
PMID:[27899579](https://pubmed.ncbi.nlm.nih.gov/27899579/)
89. Ward LD, Kellis M. HaploReg v4: systematic mining of putative causal variants, cell types, regulators and target genes for human complex traits and disease. *Nucleic Acids Res.* 2016; 44:D877–81.
<https://doi.org/10.1093/nar/gkv1340>
PMID:[26657631](https://pubmed.ncbi.nlm.nih.gov/26657631/)
90. Lu Y, Quan C, Chen H, Bo X, Zhang C. 3DSNP: a database for linking human noncoding SNPs to their three-dimensional interacting genes. *Nucleic Acids Res.* 2017; 45:D643–49.
<https://doi.org/10.1093/nar/gkw1022>
PMID:[27789693](https://pubmed.ncbi.nlm.nih.gov/27789693/)

SUPPLEMENTARY MATERIALS

Supplementary Table

Please browse Full Text version to see the data of Supplementary Table 1.

Supplementary Table 1. Information of susceptibility SNPs associated with atrial fibrillation.