Microarray analysis of long non-coding RNA expression profiles in low high-density lipoprotein cholesterol disease

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Abstract

Background: Low high-density lipoprotein cholesterol (HDL-C) disease with unknown etiology has a high prevalence in the Xinjiang Kazak population. In this study, long non-coding RNAs (lncRNAs) that might play a role in low HDL-C disease were identified.

Methods: Plasma samples from 10 eligible individuals with low HDL disease and 10 individuals with normal HDL-C levels were collected. The lncRNA profiles for 20 Xinjiang Kazak individuals were measured using microarray analysis.

Results: Differentially expressed lncRNAs and mRNAs with fold-change values not less than 1.5 and FDR-adjusted P-values less than 0.05 were screened. Bioinformatic analyses, including Gene Ontology (GO), Kyoto Encyclopedia of Genes and Genomes (KEGG), and network analyses, were used to determine relevant signaling pathways and predict potential target genes. In total, 381 lncRNAs and 370 mRNAs were differentially expressed based on microarray analysis. Compared with those in healthy individuals, several lncRNAs were upregulated or downregulated in patients with low HDL-C disease, among which TCONS_00006679 was most significantly upregulated and TCONS_00011823 was most significantly downregulated. GO and KEGG pathway analyses as well as co-expression networks of lncRNAs and mRNAs revealed that the platelet activation pathway and cardiovascular disease were associated with low HDL-C disease.

Conclusions: Potential target genes integrin beta-3 (ITGB3) and thromboxane A2 receptor (TBXA2R) were regulated by the lncRNAs AP001033.3–201 and AC068234.2–202, respectively. Both genes were associated with cardiovascular disease and were involved in the platelet activation pathway. AP001033.3–201 and AC068234.2–202 were associated with low HDL-C disease and could play a role in platelet activation in cardiovascular disease. These results reveal the potential etiology of dyslipidemia in the Xinjiang Kazakh population and lay the foundation for further validation using large sample sizes.

Keywords: Low HDL-C disease, Long non-coding RNA, Microarray analysis, Platelet activation, TBXA2R, ITGB3, Gene chip

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Background
The dyslipidemia epidemic is a public health concern affecting millions of individuals worldwide. Low-density lipoprotein cholesterol (HDL-C) disease is an important type of dyslipidemia, as demonstrated by an epidemiological study showing that HDL-C levels are inversely associated with the development of cardiovascular disease [1].

HDL-C is thought to provide protection against cardiovascular disease by regulating cholesterol efflux from peripheral tissues [2], and low HDL-C disease has been recognized as a complex, multifactorial polygenic trait regulated by multiple genes and environmental factors [3].

A previous study has shown that serum HDL-C levels have a strong genetic basis with up to 80% heritability, and detected variants explain only 10% of HDL-C variation [4]. Cholesterol ester transfer protein (CETP) [5, 6], ATP-binding cassette transporter A1 (ABCA1) [7], apolipoprotein A-1 (APOA-1) [8], and cholesterol acyltransferase (LCAT) [9] are associated with low HDL-C disease.

In a genome-wide association study, almost all screened single nucleotide polymorphisms (SNPs) were located in non-coding regions; however, the contribution of long non-coding RNAs (lncRNAs) to low HDL-C disease remains unexplored.

LncRNAs, transcripts longer than 200 nucleotides lacking coding potential [10], play crucial roles in various key biological processes and human diseases [11]. However, research on the role of lncRNAs in the development of dyslipidemia is still in its preliminary stage [12, 13].

In the present study, gene chip technology was applied to identify genome-wide differences in the lncRNA composition between the serum of healthy subjects and individuals with low HDL-C disease from a Kazakh population. Differentially expressed lncRNAs and mRNAs were examined and the relationships among these differentially expressed genes and their contributions to low HDL-C disease were analyzed.

Methods
Study population
A total of 20 subjects were recruited, including 10 subjects with low HDL-C disease and 10 healthy individuals. The inclusion criterion for subjects with low HDL-C disease was serum HDL-C level < 1.04 mmol/L [14].

Exclusion criteria were the presence of obesity, tumor, diabetes, coronary artery disease (CAD), stroke, and medication history, such as NSAIDs (such as aspirin).

Whole blood (5 mL) from each participant was collected in an EDTA-containing tube and stored at –80 °C until use.

RNA extraction, labeling, and hybridization
Total RNA (including lncRNAs) was extracted from the peripheral serum using TRIzol reagent (Invitrogen, Carlsbad, CA, USA) and purified using the NucleoSpin® RNA Purification Kit (Life Technologies Corporation, Grand Island, NY, USA) according to the manufacturer’s protocol. Spectrophotometry (NanoDrop ND-1000) was performed to determine RNA purity and concentration at OD260/280. RNA integrity was confirmed using 1% formaldehyde denaturing gel electrophoresis (Agilent Technologies, Santa Clara, CA, USA).

RNA amplification, labeling, and hybridization
The cDNA labeled with a fluorescent dye (Cy5 or Cy3-dCTP) was generated using Eberwine’s linear RNA amplification method and subsequent enzymatic reaction, as described previously [16]. The CapitalBio cRNA Amplification and Labeling Kit (CapitalBio, Beijing, China) was used to produce higher yields of labeled cDNA. The cRNA amplification and labeling procedure is depicted in Fig. 1.

Microarray imaging and data analysis
Gene Spring GX was used to analyze the lncRNA and mRNA array data, including data aggregation, normalization, and quality control. To select differentially expressed genes, fold changes of ≥1.5 and ≤ 1.5 a P-value of < 0.05 for t-tests were selected as thresholds. The Adjust Data function of Cluster 3.0 was used to perform Log2 conversion of the data, median positioning centered on genes, and hierarchical clustering with average links for further analysis. Finally, JavaTreeView (Stanford University School of Medicine, Stanford, CA, USA) was used to visualize the tree.
Construction of the coding–non-coding gene co-expression network

Based on a correlation analysis of the differentially expressed lncRNAs and mRNAs, a co-expression network (CNC network) was constructed. For each pair of genes, Pearson correlation coefficients (PCCs) were calculated and pairs with significant correlations were used to construct the network. The open source bioinformatics software Cytoscape was used to generate the lncRNA and mRNA mapping network with a PCC with an absolute value of not less than 0.90 and \( P < 0.05 \).

Cis-lncRNA prediction

Cis-acting lncRNA prediction was conducted based on close correlations with a group of expressed protein-coding genes (PCC > 0.90). Each protein-coding gene and lncRNA gene were within 10 kb of each other throughout the genome [17].

Trans-lncRNA prediction

Blat tools (Standalone BLAT v. 35 × 1 fast sequence search command line tool, downloaded from: http://hgdownload.cse.ucsc.edu/admin/exe/) were used to perform reverse transcription to compare the complete sequence of lncRNA and the 3’UTR of the corresponding co-expressed mRNA with default parameter settings.

Functional analysis of differentially expressed genes

GO and KEGG pathway analyses

The differentially expressed genes were input into KOBAS for annotation and visualization. GO was used to identify the molecular functions overrepresented in the gene profile and KEGG was used to analyze the pathways related to the genes, the

Table 1 General characteristics of the two groups

| Variable                    | Case          | Control       | \( P \) value |
|-----------------------------|---------------|---------------|--------------|
| Male/female                 | 5/5           | 5/5           | 1            |
| Age (years)                 | 29.540 ± 5.627| 29.340 ± 6.707| 1            |
| HDL-C, mmol/L               | 0.725 ± 0.047  | 1.562 ± 0.117  | < 0.001      |
| TG, mmol/L                  | 0.945 ± 0.417  | 0.821 ± 0.327  | 0.579        |
| LDL-C, mmol/L               | 2.170 ± 0.774  | 1.546 ± 0.502  | 0.052        |
| Systolic pressure, mmHg     | 121.250 ± 8.638| 125.200 ± 6.210| 0.19         |
| Diastolic pressure, mmHg    | 72.050 ± 4.425 | 73.400 ± 7.321 | 0.436        |
| BMI, kg/m²                  | 20.490 ± 2.676 | 22.400 ± 1.146 | 0.089        |

Values were presented as mean ± SD, t-test or Wilcoxon rank sum test was used to obtained the \( P \) value for continuous variables, \( P < 0.05 \) significant.
Fig. 2 Box plots summarizing the lncRNA and mRNA profiles. After normalization, the distributions of the log2-ratios of fluorescence signals in the tested samples were nearly identical among samples. Blue indicates control samples and red indicates case samples.

Fig. 3 lncRNA (a) and mRNA (b) expression in the low HDL-C disease group and healthy individuals.
significance of GO term enrichment in the differentially expressed mRNAs was denoted by $P$-value ($P < 0.05$ was considered statistically significant).

To study the enrichment of potential genes and gene products in biological processes (BP), cellular components (CC), and molecular functions (MF), a GO analysis of differentially expressed mRNAs was performed. The $P$-value was used to determine significant changes in differentially expressed genes and GO.

To understand whether the dysregulated IncRNAs are involved in the regulation of genes and related signaling pathways associated with low HDL-C disease, potential targets of IncRNAs in the database via cis- and trans-regulation were predicted.

**Statistical analysis**
Statistical analyses were performed to compare the two groups in the microarray. Fold change values and
Student’s t-tests were used to analyze the statistical significance of the microarray results. The threshold value for screening differentially expressed lncRNAs and mRNAs was a fold change of ≥1.5 (P < 0.05). The false discovery rate (FDR) was used to correct the P-value. The differentially expressed genes were analyzed using Student’s t-tests implemented in SPSS (version 22.0; SPSS, Chicago, IL, USA). Fisher’s exact test was used for the KEGG pathway analysis and to compare healthy subjects with patients, where appropriate. PCC was used for the GO analysis and for the prediction of correlations between lncRNAs and protein-coding genes. P < 0.05 was considered statistically significant.

Results
General characteristics of patients and healthy individuals
Table 1 summarizes the characteristics of all individuals. There were significant differences in HDL-C levels between the two groups (P < 0.05). No differences in sex, age, TG, LDL-C, systolic pressure, diastolic pressure, and BMI were observed between the two groups (P > 0.05).

Total RNA was extracted and purified from the blood samples. The samples were of good quality and could be used for the microarray analysis. The distributions of lncRNA and mRNA levels are shown in Fig. 2.

lncRNA and mRNA expression profiles
Differentially expressed lncRNAs
Based on the microarray data, 381 lncRNAs were differentially expressed (fold change ≥1.5, P < 0.05) between individuals with low HDL-C disease and healthy individuals, including 166 up-regulated lncRNAs and 215 down-regulated lncRNAs in the low HDL-C disease group (Fig. 3). A hierarchical clustering analysis was used to arrange specimens into groups according to expression levels (Fig. 4). The top 15 differentially expressed lncRNAs are listed in Tables 2 and 3; TCONS_00006679 was the most significantly up-regulated lncRNA (fold change: 2.9543) and TCONS_00011823 was the most significantly down-regulated lncRNA (fold change: 4.8541) (Fig. 5).

Differentially expressed mRNAs
According to the microarray results, 370 mRNAs were differentially expressed (fold change ≥1.5, P < 0.05) between the low HDL-C disease and healthy groups, using the same criteria applied to lncRNAs. Among these, 285 mRNAs were up-regulated and 85 mRNAs were down-regulated in the disease group. The top 15 differentially expressed mRNAs are listed in Tables 4 and 5. OLFM4 was the most significantly up-regulated mRNA (fold change: 6.86421, P < 0.05) and LOC102723740 was the most significantly down-regulated mRNA (fold change: 4.05710, P < 0.05).

| Table 2 | Top 15 up-regulated lncRNAs in individuals with low HDL-C disease compared with those in controls |
|---------|-------------------------------------------------|
| lncRNA ID | FC | P-value | Chromosome | Length (bp) |
| TCONS_00006679 | 2.95413 | 0.0012 | 3 | 264 |
| ENST00000531609.1 | 2.55680 | 0.0357 | 14 | 1442 |
| ENST00000568189.1 | 2.42925 | 0.0120 | 2 | 735 |
| NR_027309.2 | 2.40050 | 0.0001 | 1 | 1913 |
| TCONS_00024405 | 2.39786 | 0.0000 | 16 | 221 |
| TCONS_00012221 | 2.36027 | 0.0290 | 6 | 240 |
| ENST00000447519.1 | 2.23095 | 0.0143 | 11 | 1426 |
| TCONS_000029160 | 2.18704 | 0.0001 | 21 | 422 |
| ENST00000492716.1 | 2.16546 | 0.0133 | 19 | 574 |
| X_427643.1 | 2.10105 | 0.0021 | 3 | 605 |
| TCONS_00006679 | 2.09404 | 0.0016 | 13 | 323 |
| TCONS_00027123 | 2.05533 | 0.0064 | 19 | 1305 |
| ENST00000454187.1 | 2.02073 | 0.0047 | 9 | 266 |
| TCONS_00009834 | 2.02024 | 0.0030 | 5 | 1274 |

Values on the X and Y axes in the scatter plot are average normalized values for each group (log2-scaled). The lncRNAs shown above the upper red line and below the lower green line are those with expression differences of 1.5 fold change between the two groups; the red and green lines indicate the fold change threshold. The red points in the plot represent the significantly differentially expressed lncRNAs (fold change > 1.5, P < 0.05), while the vertical lines correspond to 1.5-fold up- and downregulation. Horizontal line indicates P = 0.05.

| Table 3 | Top 15 down-regulated lncRNAs in individuals with low HDL-C disease compared with those in controls |
|---------|-------------------------------------------------|
| lncRNA ID | FC | P-value | Chromosome | Length (bp) |
| TCONS_00011823 | 4.85413 | 0.0323 | Chr6 | 383 |
| ENST00000425176.1 | 4.42477 | 0.0316 | Chr2 | 317 |
| TCONS_00008360 | 3.70280 | 0.0017 | Chr4 | 826 |
| X_427643.1 | 3.58729 | 0.0315 | Chr5 | 260 |
| ENST00000441169.1 | 3.19771 | 0.0051 | Chr3 | 576 |
| TCONS_00011801 | 2.87082 | 0.0013 | Chr6 | 543 |
| ENST00000465596.1 | 2.83830 | 0.0013 | Chr8 | 710 |
| ENST000005064872.1 | 2.67340 | 0.0279 | Chr18 | 861 |
| TCONS_00016749 | 2.58434 | 0.0277 | Chr9 | 548 |
| TCONS_00026975 | 2.45822 | 0.0251 | Chr19 | 348 |
| NR_027146.1 | 2.43441 | 0.0001 | Chr17 | 1549 |
| uc.252- | 2.32473 | 0.0222 | Chr9 | 231 |
| ENST00000416556.1 | 2.32393 | 0.0321 | Chr13 | 376 |
| ENST00000427953.1 | 2.32227 | 0.0171 | Chr1 | 1321 |
| TCONS_00003905 | 2.31021 | 0.0085 | Chr2 | 628 |
Bioinformatics analysis and characterization of differentially expressed lncRNAs

GO and KEGG pathway and disease analyses

A GO analysis was performed using Gene Ontology (www.geneontology.org), which provides three structured networks of defined terms that describe gene product attributes. The GO results showed that many differentially expressed genes were enriched in the BP, MF and CC categories, and the top 15 enriched terms are shown in Fig. 6. These results indicated that the most significantly enriched GO terms associated with differentially expressed transcripts were ‘platelet alpha granule’ (GO:0031091, ontology: cellular component; \( P = 1.98\times10^{-14} \)), ‘wound healing’ (GO:0042060, ontology:...

Table 4: Top 15 up-regulated mRNAs in the low HDL-C disease group compared with those in controls

| Gene Symbol | FC (abs) | \( P \)-value | Chromosome |
|-------------|---------|--------------|------------|
| OLFM4       | 6.86421 | 0.0142       | chr13:53626095–53,626,154 |
| DEFA3       | 4.01720 | 0.0239       | chr8:6873570–6,873,511 |
| PFAV1       | 3.88383 | 0.0048       | chr4:74720137–74,720,196 |
| MMP8        | 3.80859 | 0.0049       | chr11:102583793–102,583,734 |
| LTF         | 3.45179 | 0.0340       | chr3:46479489–46,477,720 |
| HBA2        | 3.32376 | 0.0103       | chr16:222889–222,948 |
| HBD         | 3.04019 | 0.0123       | chr11:5255317–5,255,258 |
| OSBP2       | 2.84201 | 0.0355       | chr22:31302848–31,302,907 |
| HBB         | 2.78993 | 0.0102       | chr11:5246777–5,246,718 |
| NFIB        | 2.66136 | 0.0179       | chr9:14088270–14,088,211 |
| HBG1        | 2.63723 | 0.0469       | chr11:5269593–5,269,534 |
| HBA2        | 2.61040 | 0.0050       | chr16:223555–223,614 |
| HBA2        | 2.57408 | 0.0067       | chr16:223560–223,619 |
| BPI         | 2.54008 | 0.0222       | chr20:36965709–36,965,768 |
| CAMP        | 2.47498 | 0.0417       | chr3:48266856–48,266,915 |

Table 5: Top 15 down-regulated mRNAs in the low HDL-C disease group compared with the healthy group

| Gene Symbol | FC (abs) | \( P \)-value | Genomic Coordinates |
|-------------|---------|--------------|---------------------|
| LOC102723740| 4.05710 | 0.0236       | chr5:41284979–41,284,920 |
| LOC101928223| 3.21873 | 0.0310       | chr4:171197767–171,197,826 |
| LOC388456   | 2.58673 | 0.0303       | chr18:926149–926,090 |
| JUN         | 2.36467 | 0.0142       | chr1:59246570–59,246,511 |
| CLEC14A     | 2.30657 | 0.0482       | chr14:38724198–38,724,139 |
| FAM149B1    | 2.29307 | 0.0075       | chr10:75003877–75,003,936 |
| MDGA1       | 2.20545 | 0.0393       | chr6:37600354–37,600,295 |
| RPF1        | 2.14091 | 0.0376       | chr1:84948889–84,948,948 |
| TLE1        | 2.01499 | 0.0062       | chr9:84214882–84,214,823 |
| BNC2        | 1.98561 | 0.0147       | chr9:16419008–16,418,949 |
| GSTG2       | 1.95754 | 0.0039       | chr1:1006057346–106,057,405 |
| LOC101928043| 1.93770 | 0.0045       | chr1:22350873–223,508,14 |
| RBM44       | 1.85520 | 0.0192       | chr2:238751106–238,751,165 |
| LOC100129171| 1.84261 | 0.0328       | chr2:224854629–224,854,570 |
| OR3A2       | 1.84140 | 0.0320       | chr17:3181347–318,128,288 |
biological process; \( P = 4.14 \times 10^{-14} \), and ‘oxygen transporter activity’ (GO:0005344, ontology: molecular function, \( P = 5.66 \times 10^{-07} \)).

Additionally, a pathway analysis indicated that 23 pathways were significantly enriched among the differentially expressed transcripts (\( P < 0.05 \)) and the top 15 pathways are shown in Fig. 7a. The most significantly enriched pathway was platelet activation (KEGG ID: hsa04611, \( P = 4.49 \times 10^{-06} \)). Based on a disease enrichment analysis, we found that the most significantly enriched disease was cardiovascular disease (\( P = 3.14 \times 10^{-05} \)) (Fig. 7b).

**Construction of a co-expression network**

A co-expression network was constructed to explore the associations between IncRNAs and potential target mRNAs with differential expression in low HDL-C disease. Using PCC ≥ 0.90 and \( P < 0.05 \) as thresholds, 467 pairs of co-expressed IncRNAs and mRNAs were identified (Figs. 8 and 9).

**Long non-coding RNA target prediction**

Target genes were identified for ten IncRNAs (Table 6). The IncRNA AP001033.3–201 (ENST00000584509.1) may contribute to the trans-regulation of the protein-coding gene thromboxane A2 receptor (TBXA2R) and the IncRNA AC068234.2–202 (ENST00000575039.1) may act as a cis-regulator of the protein-coding gene integrin subunit beta 3 (ITGB3). Both target genes are involved in the platelet activation pathway and cardiovascular disease.

**Discussion**

Low HDL-C disease has the highest prevalence among the three types of dyslipidemia [18, 19] in the Xinjiang Kazakh population. Epidemiological studies have shown that HDL-C levels play a crucial role in the development of atherosclerosis and CAD [20], and low HDL-C disease has been identified as an independent traditional risk factor for CAD [21, 22]. Research has mainly focused on the protein-coding genes associated with
low HDL-C disease [5, 7, 8, 23, 24], and little is known about the underlying genetic mechanisms, including the roles of lncRNAs, which are closely involved in various important biological processes. Researchers have shown that lncRNAs are involved in the regulation of HDL metabolism [25]; HDL in patients with CAD and hypercholesterolemia could cause the abnormal expression of lncRNAs in vascular endothelial cells and have adverse effects on vascular function. A recent microarray analysis has demonstrated that the APOA1 gene cluster is regulated by the lncRNA APOA1-AS [26], which encodes APOA1, a major protein component of HDL in the plasma. Moreover, the lncRNA known as lncHR1 (HCV regulated 1) decreases lipid metabolism by repressing SREBP-1c gene expression [13]. However, the expression profiles of lncRNAs in low HDL-C disease have not been reported to date.

LncRNAs may function by altering the expression of protein-coding genes [27]. They are key regulators of gene expression, and dysregulated expression levels of certain lncRNAs are associated with a variety of diseases, including diseases arising from defects in lipid metabolism [12, 28, 29]. Generally, cholesterol biosynthesis in the liver is important for the subsequent synthesis of lipids and lipoproteins and for the correct functioning of lipid metabolism [30, 31]. Aberrant expression of lncRNAs may induce specific mRNAs to alter HDL-C levels, which would ultimately lead to low HDL-C disease.

Fig. 7 Pathway and disease analysis results (top 15) for differentially expressed mRNAs in low HDL-C disease. FDR-corrected P-values are shown.
In this study, genome-wide lncRNA and mRNA expression profiles were compared between individuals with low HDL-C disease and healthy individuals using a microarray approach and potential functions of differentially expressed loci were evaluated by GO and KEGG pathway analyses. A large number of differentially expressed lncRNA and mRNAs between low HDL-C disease and controls were identified. GO and KEGG pathway analyses showed that the differentially expressed genes were involved in a variety of cellular processes, cellular components, and molecular functions.

To elucidate the molecular mechanisms by which lncRNAs contribute to low HDL-C disease, a CNC network was further constructed by combining aberrantly expressed lncRNAs and mRNAs on the basis of lncRNA levels having an effect on the expression of flanking protein-coding genes [32]. The results of the current study indicated that multiple lncRNAs were clearly associated with mRNAs and the co-expression network could provide a strong foundation for predicting the functions of lncRNAs.

Additionally, a pathway analysis identified 23 pathways related to low HDL-C disease. Among these pathways, platelet activation (KEGG ID: hsa04611) involved 14 differentially expression genes and was the most significantly enriched pathway.

A disease enrichment analysis showed that 370 differentially expressed mRNAs were associated with many diseases, such as CAD, hematologic diseases, Bernard-
Soulier syndrome, thalassemia, systemic sclerosis, and dilated cardiomyopathy; the greatest enrichment was observed for CAD.

Based on genomic annotation, enrichment, co-expression, and target gene prediction analyses, 10 lncRNAs with target genes were identified. Based on combined GO, KEGG pathway, and disease enrichment analyses, the lncRNAs AC068234.2–202 and AP001033.3–201 were co-expressed with the target genes ITGB3 and TBXA2R.

Finally, the study results indicated that ITGB3 and TBXA2R were both differentially expressed between the two groups and were both related to the platelet activation pathway and cardiovascular disease in a GO enrichment analysis. They were also co-expressed with the lncRNAs AC068234.2–202 and AP001033.3–201 in CNC networks and were identified as potential target genes of the lncRNAs AC068234.2–202 and AP001033.3–201.

The lncRNA AP001033.3–201 was a novel transcript located on chromosome 18:9,310,522–9,334,445 reverse.
strand. The transcript had 3 exons with 5282 reported variant alleles and mapped to 93 oligo probes. The transcript was a product of the gene AP001033.3. The lncRNA AC068234.2–202 was also a novel transcript, antisense to ITGB3, located on chromosome 17:47,303,474–47,323,613 reverse strand. The transcript had 3 exons, associated with 4518 variant alleles and maps to 115 oligo probes, and the transcript was a product gene AC068234.2.

Based on a literature review, ITGB3 is an important member of the integrin family of adhesion molecules. It is located on chromosome 17:47,253,827–47,313,743 forward strand. It is associated with thrombus formation and platelet aggregation [33], abnormal platelet activation and abnormal thrombosis [34], and inflammatory responses and atherosclerosis. TBXA2R, located on chromosome 19:3,594,507–3,606,875 reverse strand, may affect platelet function and venous thrombosis [35], platelet aggregation, and the process of atherosclerosis. Usually, TBXA2R is widely distributed in the cardiovascular system. However, the distribution of TBXA2R is altered in various cardiovascular diseases and is involved in pathophysiological processes. Platelet activation plays an important physiological role in inflammation, and extensive clinical and experimental evidence supports the importance of inflammatory processes in both the initiation and development of dyslipidemia [36, 37].

Based on the above analysis, the results of this study suggest that the IncRNA AC068234.2–202 is a cis-acting regulator of ITGB3 and the IncRNA AP001033.3–201 is a trans-acting regulator of TBXA2R, contributing to the induction of platelet activation and subsequently to the pathogenesis of CAD. These results provide a foundation for further investigations of IncRNA functions, signaling pathways, and roles in low HDL-C disease.

To conclude, the expression profiles of IncRNAs and mRNAs in low HDL-C disease were determined using a microarray approach, and 381 IncRNAs and 370 mRNAs were identified. An IncRNA–mRNA co-expression network and predicted network were constructed and combined with the results of GO, KEGG pathway, and disease enrichment analyses. Results indicated that the IncRNA AP001033.3–201 is a trans-regulatory element for the protein-coding gene TBXA2R and the IncRNA AC068234.2–202 is a cis-regulatory element for the protein-coding gene ITGB3. Both of these IncRNAs were involved in platelet activation and CAD. Based on these results and microarray data, the results suggest that platelet activation plays an important role in the occurrence and development of low HDL-C disease.

Moreover, Zhong et al. reported that both high and low HDL-C levels (ranging from 22 mg/dL to 97 mg/dL) were associated with an increased risk of mortality from CAD [38]. The IncRNAs identified in this study provide insights into the potential molecular mechanisms underlying the effects of low HDL-C levels. However, further research is needed to confirm the functions of the candidate differentially expressed IncRNAs and their potential regulatory relationships in low HDL-C disease.

### Study strengths and limitations

A large number of differentially expressed IncRNAs and mRNAs in low HDL-C disease were successfully screened in a Kazak population and potential target genes were predicted by microarray and bioinformatic analyses. However, the study had certain limitations. First, it was based on the results of bioinformatics analyses, and further biological experiments, such as qPCR on an independent cohort, are needed to verify the findings. Second, the cross-sectional study design could not reveal whether changes in IncRNA and mRNA expression were primary or secondary changes in the disease. Finally, it is important to acknowledge that the small study cohort might contribute to a reduced statistical power, to some extent.

### Table 6 Potential target gene prediction for differentially expressed IncRNAs

| IncRNA    | mRNA         | Correlation | P-value   | IncRNA    | mRNA         |
|-----------|--------------|-------------|-----------|-----------|--------------|
| XR_429369.1 | LOC101927780 | 0.9669      | 4.05165E-12 | down     | down         |
| ENST00000602325.1 | PTGS1       | 0.9436      | 4.45907E-10 | up       | up           |
| TCONS_00008319 | LOC101928223 | 0.9615      | 1.51796E-11 | down     | down         |
| uc.252-  | BNC2         | 0.9339      | 1.79422E-09 | down     | down         |
| ENST00000508021.1 | SNCA       | 0.9732      | 6.16618E-13 | up       | up           |
| ENST00000564872.1 | LOC388456  | 0.9681      | 2.9301E-12  | down     | down         |
| ENST00000558838.1 | PCSK6      | 0.9894      | 0          | up       | up           |
| ENST00000575039.1 | ITGB3      | 0.9517      | 1.14575E-10 | up       | up           |
| ENST00000584509.1 | TBXA2R     | 0.9206      | 8.85915E-09 | up       | up           |
| ENST00000443574.1 | HLA-DQB1   | 0.9378      | 1.05879E-09 | up       | up           |
Conclusions
In this study, IncRNA and mRNA expression profiles in low HDL-C disease were screened in a Kazakh population, and a large number of IncRNAs and mRNAs were found to be differentially expressed. GO, KEGG pathway, and disease analyses suggested that the IncRNAs AC068234.2–202 and AP001033.3–201 may regulate the genes ITGB3 and TXB2AR, respectively. These loci are involved in platelet activation and the pathogenesis of CAD. This microarray analysis clarified the expression patterns of IncRNAs in low HDL-C disease and laid the foundation for future functional and mechanistic studies of IncRNAs associated with low HDL-C disease.

Furthermore, key candidate genes associated with low HDL-C disease were identified, providing insights into the underlying genetic basis as well as a basis for follow-up research, including functional analyses of IncRNAs and studies of the roles of signaling pathways in the HDL-C disease in the Kazakh population in Xinjiang. In addition, the results of this study improve our understanding of the etiology of both low HDL-C and CAD as well as the relationship between the diseases.

Abbreviations
HDL-C: High-density lipoprotein cholesterol; CETP: Cholesterol ester transfer protein; ABCA1: ATP-binding cassette transporter A1; APOA-1: Apolipoprotein A-1; IncRNAs: Long non-coding RNAs; CAD: Coronary artery disease; ITGB3: Integrin beta-3; TXB2AR: Thromboxane A2 receptor

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Authors’ contributions
Conceptualization, Shuxia Guo and Heng Guo; methodology, Xianghui Zhang; software, Yizhong Yan; validation, Jiaolong Ma and Yu Li; formal analysis, Xinping Wang; investigation, Shuxia Guo; resources, Haixia Wang; data curation, Yunhua Hu; writing—original draft preparation, Xinping Wang; writing—review and editing, Jia He; visualization, Xinping Wang; supervision, Rulin Ma; project administration, Shuxia Guo. The author(s) read and approved the final manuscript.

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Availability of data and materials
The datasets used and/or analyzed during the current study are available from the corresponding author on reasonable request.

Ethics approval and consent to participate
The present study was approved by the Institutional Ethics Review Board at the First Affiliated Hospital of Shihze University School of Medicine (ERB No. SHZ2010LL01). All participants provided written informed consent before the study began.

Consent for publication
Not applicable.

Competing interests
The authors declare that they have no competing interests.

References
1. Collaboration. TERF: Major lipids, apolipoproteins, and risk of vascular disease. 2009.
2. Chang FJ, Yuan HY, Hu XX, Ou ZJ, Fu L, Lin ZB, Wang ZP, Wang SM, Zhou L, Xu YQ. High-density lipoprotein from patients with valvular heart disease uncouples endothelial nitric oxide synthase. J Mol Cell Cardiol. 2014;74:209–19.
3. Wang X, Guo H, Yu L, Wang H, Jia H, Mu L, Hu Y, Ma J, Yan Y, Li S. Interactions among genes involved in reverse cholesterol transport and in the response to environmental factors in dyslipidemia in subjects from the Xinjiang rural area. PLoS One. 2018;13:e0196042.
4. Boes E, Coassin S, Kolletts B, Held IM, Koenenberg F. Genetic-epidemiological evidence on genes associated with HDL cholesterol levels: a systematic in-depth review. Exp Gerontol. 2009;44:136–60.
5. Niesor EJ, Kallend D, Bentley D, Kastelijn JJP, Kees Hovingh G, Stroes ESG. Treatment of low HDL-C subjects with the CETP modulator dalteparin increases plasma campesterol only in those without ABCA1 and/or APOA1 mutations. Lipids. 2014;49:1245–9.
6. Jing G, Mao YM, Cong HL, Yi L, Cui RZ. Relationship between cholesteryl ester transfer protein gene -629C→A mutations with HDL-C levels and coronary heart disease. Chin Med Abstracts Intern Med. 2013;4:399–405.
7. Yao M-H, Guo H, He J, Yan Y-Z, Ma R-L. Interactions of six SNPs in ABCA1 gene and obesity in low HDL-C disease in Kazakh of China. Int J Environ Res Public Health. 2016;13:2176.
8. Wang X, He J, Guo H, Mu L, Hu Y, Ma J, Yan Y, Ma R, Li S, Ding Y, et al. Interactions of six SNPs in APOA1 gene and types of obesity on low HDL-C disease in Xinjiang pastoral area of China. Lipids Health Dis. 2017;16:187.
9. Holleboom AG, Kuivenhoven JA, Peelman F, Schimmel AW. High prevalence of mutations in LCAT in patients with low HDL cholesterol levels in the Netherlands: Identification and characterization of eight novel mutations. Hum Mutat. 2011;32(11):1209–8.
10. Mercer TR, Dinger ME, Mattick JS. Long non-coding RNAs: insights into functions. Nat Rev Genet. 2009;10:155–9.
11. Zhang D, Wang B, Ma M, Yu K, Zhang Q, Zhang X. IncRNA HOTAR protects myocardial infarction rat by sponging miR-519d-3p. J Cardiovasc Transl Res. 2019;12:171–83.
12. Lang-Ouellette D, Richard T, Morin P. Mammalian hibernation and regulation of lipid metabolism: a focus on non-coding RNAs. Biochemistry. 2014;79:161–71.
13. Duan L, Min C, Niu Y, Chi X, Liu X, Fan J, Fan H, Chang Y, Wei Y. Identification of a novel human long non-coding RNA that regulates hepatic lipid metabolism by inhibiting SREBP-1c. Int J Biol Sci. 2017;13:349–57.
14. Zhu JR, Gao RL, Zhao SP, Lu GP, Zhao D, Li JI. Guidelines for the prevention and treatment of dyslipidemia in Chinese adults (2016 revision). Chin J Circ. 2016;41:155–35.
15. Ørom UA, Derrien T, Beringer M, Gumireddy K, Gardini A, Bussotti G, Lai F, Zytnicki M, Notredame C, Huang Q. Long noncoding RNAs with enhancer-like function in human cells; 2010.
16. Patterson TA, Lobenhofer EK, Fulmer-Smentek SB, Collins PJ, Chu TM, Bao W, Fang H, Kawasaki ES, Hager J, Tikhonova IR, et al. Performance comparison of one-color and two-color platforms within the Microarray Quality Control (MAQC) project. Nat Biotechnol. 2006;24:1140–50.
17. Je H, Ohak M, Bogu GK, Stanton LW, Johnson R, Lipovich L. Genomewide computational identification and manual annotation of human long noncoding RNA genes. RNA. 2010;16:1478–87.
18. Joshi SR, Anjana RM, Deepa M, Pradeepa R, Bansali A, Dhandania VK, Joshi PP, Unnikrishnan R, Nirmal E, Subashini R, et al. Prevalence of dyslipidemia in urban and rural India: the ICMR-INDIAB study. PLoS One. 2014;9:e96808.
19. Guo SX, Ma RL, Guo H, Ding YS, Liu JM, Zhang M, Zhang JY, Xu SZ, Li SG, Rui DS, et al. Epidemiological analysis of dyslipidemia in adults of three ethnicities in Xinjiang, China. Genet Mol Res. 2014;13:2385–93.
20. Peters SAE, Ruitter HMD, Bols TL, Moons KGM. Improvements in risk stratification for the occurrence of cardiovascular disease by imaging subclinical atherosclerosis: a systematic review. Heart. 2011;98:177–84.
21. Castelli W. Cholesterol and lipids in the risk of coronary artery disease—the Framingham heart study. Can J Cardiol. 1988;4:Suppl A:SA–10A.

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22. Ingelsson E, Schaefer EJ, Contois JH, McNamara JR, Sullivan L, Keyes MJ, Pencina MJ, Schoonmaker C, Wilson PW, D’Agostino RB, Vasan RS. Clinical utility of different lipid measures for prediction of coronary heart disease in men and women. JAMA. 2007;298:776–85.

23. Klos RL, Sing CF, Boerwinkle E, Hamon SC, Rea TJ, Clark A, Fornage M, Hixson JE. Consistent effects of genes involved in reverse cholesterol transport on plasma lipid and apolipoprotein levels in CARDIA participants. Arterioscler Thromb Vasc Biol. 2006;26:1828–36.

24. Ma XY, Liu JP, Song ZY. Associations of the ATP-binding cassette transporter A1 R219k polymorphism with HDL-C level and coronary artery disease risk: a meta-analysis. Atherosclerosis. 2011;215:428–34.

25. Liu X, Wang TT, Li Y, Shi MM, Li HM, Yuan HK, Mo ZW, Chen J, Zhang B, Chen YX, et al. High density lipoprotein from coronary artery disease patients caused abnormal expression of long non-coding RNAs in vascular endothelial cells. Biochem Biophys Res Commun. 2017;487:552–9.

26. Halley P, Kadakkuzha BM, Faghhihi MA, Magistri M, Zeier Z. Regulation of the apolipoprotein gene cluster by a long noncoding RNA. Cell Rep. 2014;6(1):222–30.

27. Pinn JL, Chang HY. Genome regulation by long noncoding RNAs. Annu Rev Biochem. 2012;81:145–66.

28. Smolle E, Haybaeck J. Non-coding RNAs and lipid metabolism. Int J Mol Sci. 2014;15:13494–513.

29. Li P, Ruan X, Yang L, Kiesewetter K, Zhao Y, Luo H, Chen Y, Gucek M, Zhu J, Cao H. A liver-enriched long non-coding RNA, lncLSTR, regulates systemic lipid metabolism in mice. Cell Metab. 2015;21:455–67.

30. Zhao C, Dahlman-Wright K. Liver X receptor in cholesterol metabolism. J Endocrinol. 2010;204:233–40.

31. Min HK, Kapoor A, Fuchs M, Meshahi F, Zhou H, Maher J, Kellum J, Warnick R, Contos MJ, Sanyal AJ. Increased hepatic synthesis and dysregulation of cholesterol metabolism is associated with the severity of nonalcoholic fatty liver disease. Cell Metab. 2012;15:666–74.

32. Wang L, Fu D, Qiu Y, Xing X, Xu F, Han C, Xu X, Wei Z, Zhang Z, Ge J, et al. Genome-wide screening and identification of long noncoding RNAs and their interaction with protein coding RNAs in bladder urothelial cell carcinoma. Cancer Lett. 2014;349:77–86.

33. Papp E, Havasi V, Bene J, Komlosi K, Coopf L, Magyar E, Feher C, Feher G, Horvath B, Manton Z, et al. Glycoprotein IIa gene (P2Y1) polymorphism and aspirin resistance: is there any correlation? Ann Pharmacother. 2002;36:1013–8.

34. Morgan EA, Schneider JG, Baron TE, Uluçkan O, Heller E, Hurchla M, Deng H, Floyd D, Berdy A, Prior J, et al. Dissection of platelet and myeloid cell defects by conditional targeting of the beta3-integrin subunit. FASEB J. 2010;24:1117–27.

35. Zhao J, Fu Y, Yang W, Yan J, Zhao J, Chen S, Xia W, Thorbomoxane A2 receptor polymorphism in association with cerebral infarction and its regulation on platelet function. Curr Neurovasc Res. 2015;12:15–24.

36. Elkind MSV. Inflammation, Atherosclerosis, and Stroke. Neurologist. 2006;12:140–8.

37. Kontush A, Chapman MJ. Functionally defective high-density lipoprotein: a new therapeutic target at the crossroads of dyslipidemia, inflammation, and atherosclerosis. Pharmacol Rev. 2006;58:342.

38. Zhong GC, Huang SQ, Peng Y, Wan L, Wu YQ, Hu TY, Hu JJ, Hao FB. HDL-C is associated with mortality from all causes, cardiovascular disease and cancer in a J-shaped dose-response fashion: a pooled analysis of 37 prospective cohort studies. Eur J Prev Cardiol. 2020;27(1):1187–203.

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