

Identification of critical microRNAs in gastrointestinal stromal tumor patients treated with Imatinib

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Gastrointestinal stromal tumors (GISTs) are the most common mesenchymal neoplasms of the gastrointestinal tract. Imatinib mesylate was considered a breakthrough drug in clinical treatment of GIST, but GIST patients showed resistance to it. We aimed to identify critical microRNAs (miRNAs) related to imatinib resistance in imatinib-treated GIST patients. Microarray datasets under accession numbers GSE63159 and GSE45901 were downloaded from the Gene Expression Omnibus (GEO) database. The differentially expressed miRNAs (DEMs) related to imatinib resistance were identified. GO function and KEGG pathway enrichment analyses were performed, and lncRNA-miRNA-target gene regulatory networks were constructed. Finally, the critical miRNAs and their target genes related to imatinib resistance or sensitivity were identified. A total of 20 DEMs in the GSE63159 dataset (7 significantly up-regulated and 13 down-regulated) and 23 DEMs in the GSE45901 dataset (8 up-regulated and 15 down-regulated) were identified. Five critical miRNAs and 109 target genes were identified in the lncRNA-miRNA-target gene regulatory networks. GO function and KEGG pathway enrichment analysis showed that target genes of DEMs were mainly involved in several signaling pathways, such as focal adhesion and the GnRH signaling pathway. From the five miRNAs, the overexpression of *hsa-miR-28-5p* and *hsa-miR-125a-5p* had significant correlation to imatinib resistance or imatinib sensitivity in GIST patients. In addition, *Hsa-miR-28-5p* and *hsa-miR-125a-5p* may be involved in the development and progression of GIST, and they may serve as prognostic markers for imatinib-response in GIST patients.

Key words: gastrointestinal stromal tumors, microRNAs, imatinib mesylate, RNA, long noncoding, focal adhesions, gonadotropin-releasing hormone

Gastrointestinal stromal tumors (GIST) are the most common mesenchymal neoplasm of the gastrointestinal tract, with an incidence of 1–2 cases/100,000 individuals/year worldwide [1]. Most GIST patients express mutation of two key oncogenic genes, v-kit Hardy-Zuckerman 4 feline sarcoma viral oncogene homolog (*KIT*) or platelet-derived growth factor receptor α (*PDGFRA*) [2, 3]. Today, imatinib mesylate is specifically applied in clinical settings to target *KIT* and *PDGFRA* encoded proteins, and it is considered a breakthrough drug in GIST treatment [4, 5]. Although the clinical application of imatinib has improved outcomes for patients, drug resistance still remains an urgent challenge for GIST therapy.

microRNAs (miRNAs) are small noncoding RNAs with 22 nucleotides and they control tumor cell growth by regulating the expression of multiple gene products and the function of cellular pathways [6]. It is recently reported that miRNAs

promote or inhibit tumor growth effects in several kinds of cancers and they have therefore been identified as targets for cancer therapy, diagnosis, and prognosis [7, 8]. Most studies on GIST focus on the effects of miRNA expression on tumorigenesis [9], overall patient survival [10] or *KIT* and *PDGFRA* mutations [11]. Although miRNAs have been reported as vital regulators in *KIT* and *PDGFRA* expression, the functional role of miRNAs in the imatinib-response in GIST patients has rarely been reported [12].

In 2014, Akcakaya et al. [13] performed analysis of miRNA expression profiles to investigate the miRNA expression signatures related to the imatinib-response and *KIT* mutational status in GIST patients. They found that *miR-125a-5p* and its target gene, tyrosine-protein phosphatase non-receptor type (*PTPN18*), played an important role in imatinib resistance in GIST cells. Herein, we used the microarray data deposited by Akcakaya et al. to identify critical miRNAs, target genes, and

potential regulatory pathways related to imatinib resistance in GIST samples. In addition, Gene Ontology (GO) function and Kyoto Encyclopedia of Genes and Genomes (KEGG) pathway enrichment analyses were performed. The related lncRNAs of miRNAs were searched, and a competing endogenous RNA (ceRNA) regulatory network was constructed. The objective of this study was to identify critical miRNAs related to imatinib resistance which could serve as important therapeutic targets or prognostic biomarkers in GIST.

Materials and methods

Data preprocessing and differentially expressed miRNA (DEM) screening. Two microarray datasets under accession numbers GSE63159 and GSE45901 [13], including 34 and 17 GIST human samples respectively, were downloaded from the NCBI (National Center for Biotechnology Information) GEO (Gene Expression Omnibus) database (<https://www.ncbi.nlm.nih.gov/>). GSE63159 included 19 GIST patient samples which were treated with imatinib and 15 untreated samples and GSE45901 included 17 GIST samples treated with imatinib. Seven of these samples were imatinib-resistant and ten were imatinib-sensitive. The two microarray datasets were analyzed based on the GPL10656 Affymetrix Human Gene 1.0 ST Array platform.

The Limma package [14] in R software (version 3.4.1) normalized the gene expression profile. The t-test (<http://127.0.0.1:26738/library/stats/html/t.test.html>) and Wilcoxon rank sum test (<http://127.0.0.1:26738/library/stats/html/wilcox.test.html>) then screened the DEMs related to GIST progression in R software (3.4.1 version). DEMs in the GSE63159 dataset were screened in the imatinib-treated GIST samples and compared to untreated samples. DEMs in the GSE45901 dataset were then screened from imatinib-sensitive GIST samples and compared to imatinib-resistant samples. Therefore, the DEMs with the cut-off criteria of a logFC greater than 0.5 and p-value less than 0.05 were selected. Finally, we used the intersection of the DEMs from the t-test and Wilcoxon-test methods, and Venn diagrams visualized the results.

Bidirectional hierarchical clustering analysis. After microarray data from the GSE63159 and GSE45901 datasets were preprocessed, the expression values of DEMs in the two groups were extracted and converted into the expression matrix using the Affy package [15]. The pheatmap package [16] in R software (3.4.1 version) analyzed the expression matrix values based on the Euclidean distance [17]. Bidirectional hierarchical clustering analysis was then performed and gene expression values were displayed in heat maps [18, 19]. Similar gene expression values could be clustered together based on the bidirectional hierarchical clustering analysis.

Construction of miRNA co-expression network and miRNA-target genes network. The R 3.4.1 software (<http://127.0.0.1:20874/library/stats/html/cor.html>) calcu-

lated the Pearson correlation coefficient between two DEMs obtained from the GSE63159 dataset; resultant miRNAs with a Pearson correlation coefficient greater than 0.8 were obtained. The miRNA-miRNA co-expression network was constructed. Target genes that were co-expressed with miRNAs were searched in three miRNA-target gene databases: miRanda [20] (<http://www.microrna.org/microrna/home.do>), miRTarBase 2016 [21] (<http://mirtarbase.mbc.nctu.edu.tw/>), and TargetScan release 7.1 [22] (<http://www.targetscan.org/>). We found intersections between the miRNAs and target genes in the three databases and constructed the regulatory network. Finally, the complex miRNA-target gene network was visualized by Cytoscape 3.4.0 software [23] (<http://www.cytoscape.org/>).

GO and KEGG pathway analysis. DAVID 6.8 performed the GO function and KEGG pathway enrichment analysis based on the GSE63159 dataset. A p-value less than 0.05 was the threshold, and the formula for calculating the p-value is as follow:

$$p = 1 - \sum_{i=0}^{H-1} \frac{\binom{M}{i} \binom{N-M}{K-i}}{\binom{N}{K}}$$

where N is the number of genes that have GO functional annotation, K is the number of DEMs in N , and M is the number of gene corresponding to a special GO function node.

Predicting the risk lncRNAs of miRNAs. The risk lncRNAs regulated by the five DEMs obtained from the GSE63159 dataset (*hsa-miR-324-5p*, *hsa-miR-342-3p*, *hsa-miR-28-5p*, *hsa-miR-125a*, and *hsa-miR-320a*) were searched using the starBase V2.0 database which is derived from 37 independent research centers, and it contains more than 6000 samples from 14 types of cancers. The interactions between lncRNAs, circRNAs, proteins and miRNAs were analyzed using this database. The relationship between miRNA-target genes and the lncRNA-miRNA-target gene network, also called the ceRNA network, was constructed and Cytoscape3.4.0 (<http://www.cytoscape.org/>) visualized the regulatory networks.

Analysis of critical genes related to imatinib resistance. The important DEMs related to imatinib resistance or imatinib sensitivity were analyzed based on the GSE45901 dataset. After integrating the DEMs found in the GSE63159 dataset, several critical DEMs related to imatinib-response were obtained. By combining these DEMs with the miRNAs-target gene regulatory network, we performed the GO function and pathway enrichment analysis to further explore the key functions of these critical DEMs.

Results

Data preprocessing and DEMs screening. The microarray data in the GSE63159 and GSE45901 datasets were normalized and the results are shown in Figure 1 (A and B).

After data preprocessing, a total of 21 DEMs from each set in the imatinib-treated samples were found to be significantly different to those in the untreated samples. This was determined by the t-test and Wilcox-test based on the GSE63159 dataset; as also were the significantly different 25 and 37

DEMs in the GSE45901 dataset for the imatinib-sensitive samples compared to the imatinib-resistant samples. After taking the intersection of these DEMs, a total of 20 DEMs were obtained in the GSE63159 (7 significantly up-regulated and 13 significantly down-regulated) and 23 DEMs in the

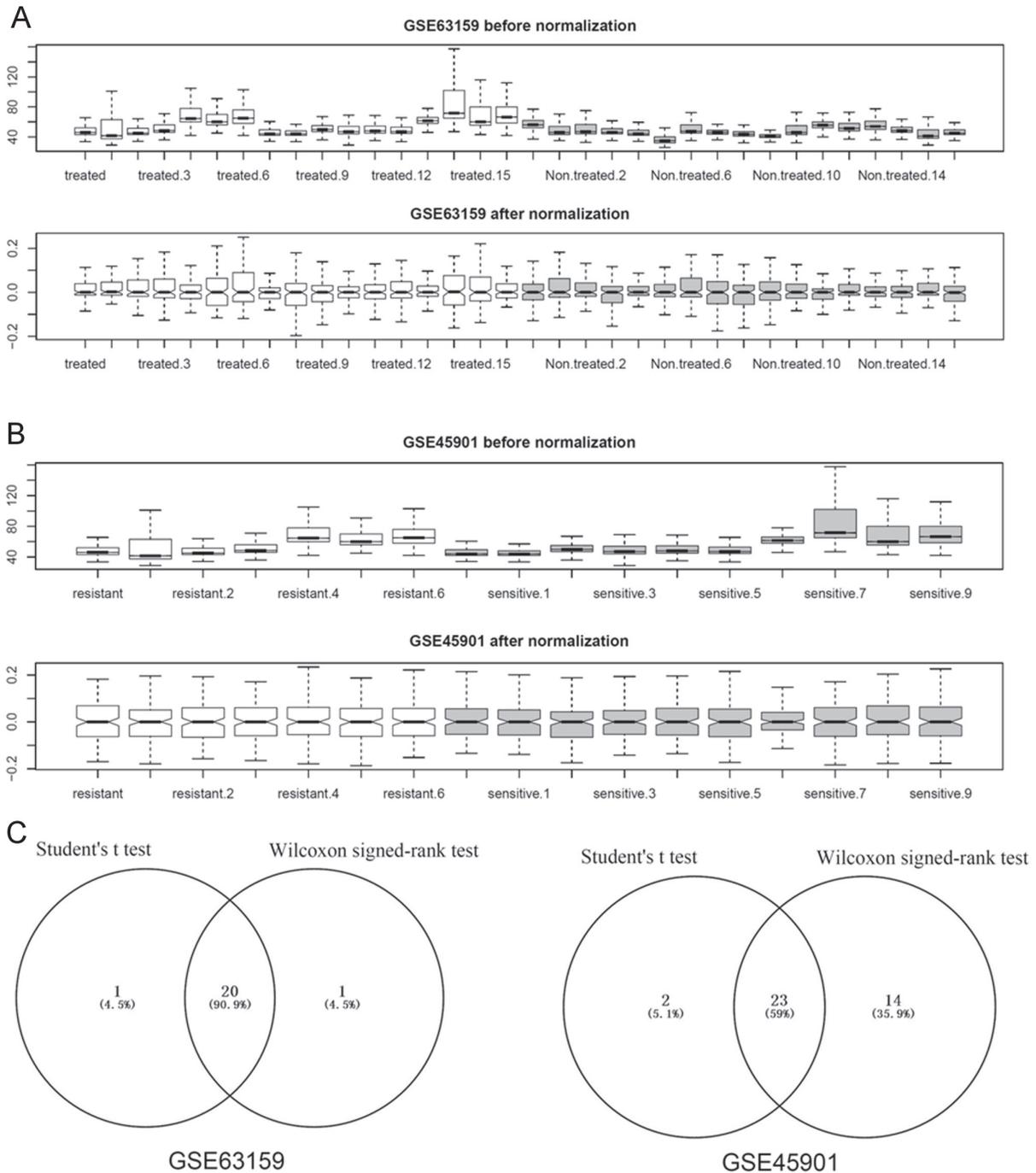


Figure 1. The normalized results of the GSE63159 and GSE45901 datasets. A) The data normalization results from the GSE63159 dataset; B) The data normalization results from the GSE45901 dataset. The white and gray boxes represent imatinib-treated gastrointestinal stromal tumor (GIST) samples and untreated GIST samples, respectively. C) The Venn diagrams show the differentially expressed miRNAs (DEMs) in the GSE63159 and GSE45901 datasets which were examined by t-test and Wilcox-test methods.

GSE45901 dataset (8 significantly up-regulated and 15 down-regulated). The results are shown in the Figure 1C Venn diagram.

Heat map of bidirectional hierarchical clustering analysis. After screening the DEMs in the GSE63159 and GSE45901 datasets, the gene expression values produced a bidirectional hierarchical clustering heat map (Figure 2). The results indicated that DEMs found in the GIST samples could be divided into different groups according to gene expression values.

miRNA-target genes regulatory network construction. DEMs in GSE63159 with a correlation coefficient >0.8 were identified. The miRNA-miRNA co-expression network was constructed (Figure 3A). The network included 19 DEMs (5 up-regulated and 14 down-regulated) and 29 edges. These relationships revealed a total of 9 miRNA pairs with negative correlation and 20 pairs with positive correlation.

Based on the results from three datasets, the miRanda, miRTarBase 2016, and TargetScan release 7.1, 52,664, 2487, and 6193 miRNA-target gene pairs were screened, respectively (Figure 3B). In these relationship pairs, 169 miRNA-target gene pairs existed in the three databases. The miRNA-target gene regulatory network was then constructed, including 5 critical miRNAs (*hsa-miR-324-5p*, *hsa-miR-342-3p*, *hsa-miR-28-5p*, *hsa-miR-125a* and *hsa-miR-320a*) and 109 target genes (Figure 3C).

GO function and KEGG pathway enrichment analysis. For further insight into the function of the 109 target genes in the regulatory network, GO function analysis and KEGG pathway enrichment analysis were performed (Figure 4, Table 1). The top twenty functions of these target genes included regulation of cellular protein metabolic processes, protein kinase cascades and cell motion.

KEGG pathway enrichment analysis was performed for these target genes. The target genes were mainly enriched in focal adhesion, the MAPK signaling pathway, neurotrophin signaling pathway, T cell receptor signaling pathway, axon guidance, leukocyte transendothelial migration, chemokine signaling pathway, adherens junction and the Fc epsilon RI signaling pathway.

Analysis of ceRNA network. A total of 14 lncRNAs were related to *hsa-miR-125a-5p* expression, including ciliary rootlet coiled-coil, rootletin pseudogene 2 (*CROCCP2*), chondroitin sulfate proteoglycan 4 pseudogene 5 (*CSPG4P5*), and Digeorge syndrome critical region gene 5 (*DGCR5*). In addition, the ceRNA network was constructed. It integrated the relationship of lncRNAs, *hsa-miR-125a-5p* and target genes. The ceRNAs network in Figure 5A contained 82 nodes (including an miRNA, 14 lncRNAs, and 67 target genes) and 81 edges (including 14 pair lncRNAs-target gene relationships and 67 pair miRNAs-target gene relationships).

GO function and KEGG pathway analysis was performed for the target genes in the ceRNA network. The top 18 functions were involved in three biological processes, eight cell components and seven molecular functions (Figure 5B, Table 2). The biological processes were related to regulation of transcription, DNA-dependent regulation of RNA metabolic processes and intracellular signaling cascades. The cell component relationships included non-membrane-bounded organelles, intracellular non-membrane-bounded organelles and the membrane-enclosed lumen. The molecular functions were related to metal ion binding, cation binding and ion binding. Furthermore, the target genes were mainly enriched in focal adhesion, GnRH signaling pathway, T cell receptor signaling pathway and leukocyte trans-endothelial migration.

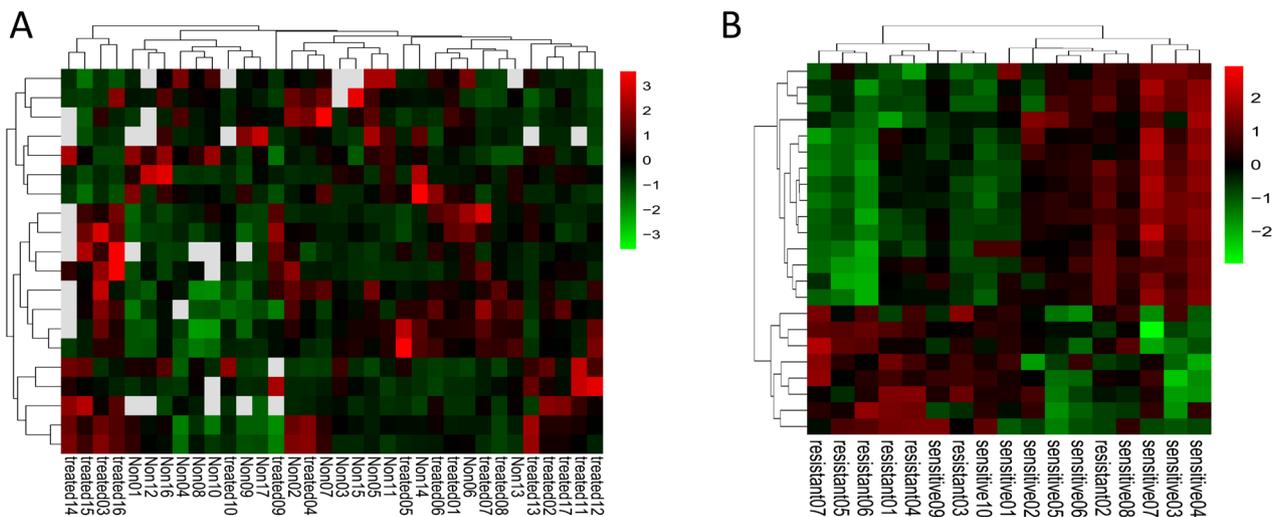


Figure 2. The bidirectional hierarchical clustering heat maps based on the expression value of differentially expressed miRNAs (DEMs) in the GSE63159 (A) and GSE45901 (B) datasets.

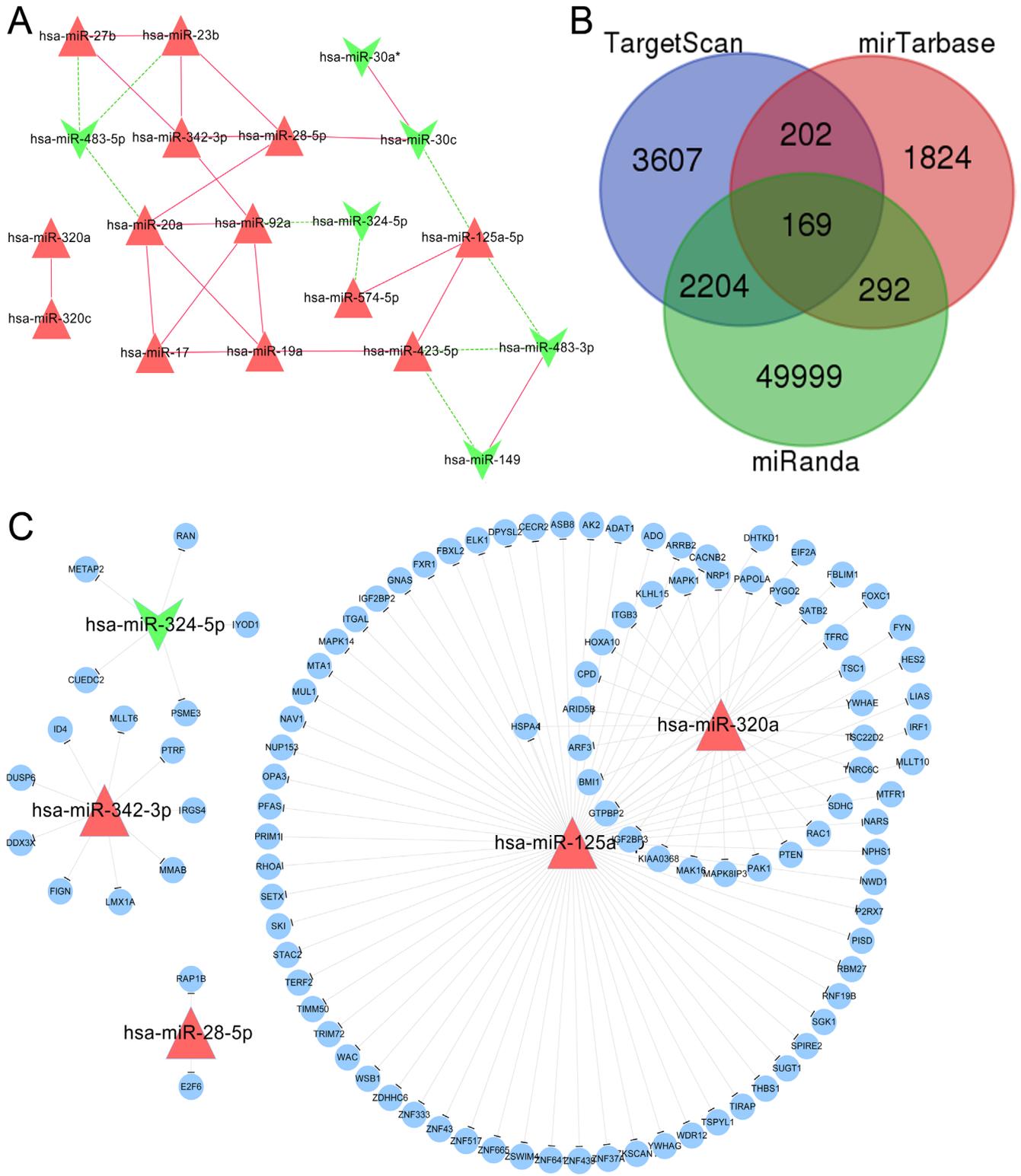


Figure 3. Construction of miRNA co-expression network and miRNA-target gene network. A) The regulatory network of differentially expressed miRNAs (DEMs) in the GSE63159 dataset. Green triangles and red triangles represent significantly down-regulated and up-regulated miRNAs, respectively. B) The Venn diagrams of target genes in three databases, miRanda, miRTarBase, and TargetScan. C) Regulatory network of miRNAs target genes; green triangles and red triangles represent significantly down-regulated and up-regulated miRNA, respectively; blue circular nodes represent target genes.

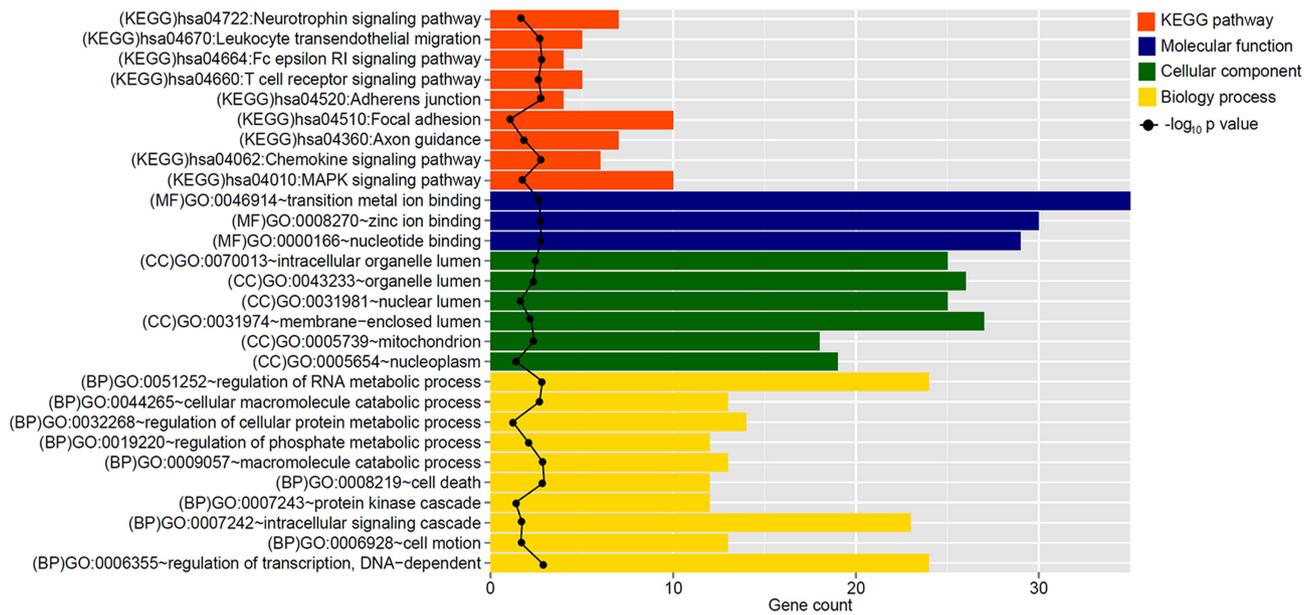


Figure 4. The Gene Ontology (GO) function and Kyoto Encyclopedia of Genes and Genomes (KEGG) pathway enrichment analysis for target genes of miRNAs. Red, blue, green, and yellow represent the KEGG pathway, MF, CC, and BP, respectively; the black line represents the $-\log_{10}$ p-value.

Table 1. GO and KEGG pathways for target genes in miRNA-target gene regulation network.

Category	Term	Count	p-value
Biology Process	GO:0032268~regulation of cellular protein metabolic process	14	2.48E-04
	GO:0007243~protein kinase cascade	12	3.86E-04
	GO:0006928~cell motion	13	8.94E-04
	GO:0007242~intracellular signaling cascade	23	0.001149
	GO:0019220~regulation of phosphate metabolic process	12	0.003399
	GO:0010605~negative regulation of macromolecule metabolic process	15	0.004701
	GO:0044265~cellular macromolecule catabolic process	13	0.024239
	GO:0006355~regulation of transcription, DNA-dependent	24	0.031822
	GO:0009057~macromolecule catabolic process	13	0.03968
	GO:0051252~regulation of RNA metabolic process	24	0.039732
Cellular Component	GO:0005654~nucleoplasm	19	4.48E-04
	GO:0031981~nuclear lumen	25	0.001003
	GO:0031974~membrane-enclosed lumen	27	0.006323
	GO:0043233~organelle lumen	26	0.009575
	GO:0005739~mitochondrion	18	0.010373
	GO:0070013~intracellular organelle lumen	25	0.013847
Molecular Function	GO:0046914~transition metal ion binding	35	0.028235
	GO:0008270~zinc ion binding	30	0.031614
	GO:000166~nucleotide binding	29	0.037275
KEGG pathway	hsa04510:Focal adhesion	10	1.92E-04
	hsa04010:MAPK signaling pathway	10	0.001531
	hsa04722:Neurotrophin signaling pathway	7	0.001674
	hsa04360:Axon guidance	7	0.002049
	hsa04660:T cell receptor signaling pathway	5	0.0254
	hsa04670:Leukocyte transendothelial migration	5	0.033718
	hsa04062:Chemokine signaling pathway	6	0.043995
	hsa04520:Adherens junction	4	0.045564
hsa04664:Fc epsilon RI signaling pathway	4	0.047047	

Table 2. GO and KEGG pathways for target genes in the ceRNA regulation network.

Category	Term	Count	p-value
Biology Process	GO:0006355~regulation of transcription, DNA-dependent	15	0.009821
	GO:0051252~regulation of RNA metabolic process	15	0.011867
	GO:0007242~intracellular signaling cascade	12	0.011145
Cellular Component	GO:0043228~non-membrane-bounded organelle	17	0.009949
	GO:0043232~intracellular non-membrane-bounded organelle	17	0.009949
	GO:0031974~membrane-enclosed lumen	16	9.09E-04
	GO:0043233~organelle lumen	15	0.002327
	GO:0031981~nuclear lumen	14	8.61E-04
	GO:0070013~intracellular organelle lumen	14	0.005489
	GO:0005654~nucleoplasm	11	6.69E-04
	GO:0005739~mitochondrion	11	0.003246
Molecular Function	GO:0046872~metal ion binding	31	4.10E-04
	GO:0043169~cation binding	31	4.91E-04
	GO:0043167~ion binding	31	6.49E-04
	GO:0046914~transition metal ion binding	27	1.80E-05
	GO:0008270~zinc ion binding	24	2.70E-05
	GO:0003677~DNA binding	17	0.028924
KEGG pathway	GO:0000166~nucleotide binding	16	0.042922
	hsa04510:Focal adhesion	4	0.046024
	hsa04912:GnRH signaling pathway	3	0.047158
	hsa04660:T cell receptor signaling pathway	3	0.04847
	hsa04670:Leukocyte transendothelial migration	3	0.04985

Discussion

In this study, we identified the *hsa-miR-28-5p* and *hsa-miR-125a-5p* were associated with imatinib resistance or imatinib sensitivity in GIST patients. GO function and KEGG pathway analysis showed that the two miRNAs regulated GIST progression through focal adhesion and the GnRH signaling pathway.

Previous studies have revealed that miRNAs play an important role in tumor pathogenesis, invasion and drug resistance in cancers [24–26], and Akakaya P [13] showed that over-expression of *hsa-miR-125a-5p* was associated with imatinib resistance in GIST patients; the potential mechanism is that *hsa-miR-125a-5p* can increase cell viability by down-regulating the expression level of *PTPN18* in tumor cells. Zheng et al. [27] showed that down-regulation of several miRNAs, including *hsa-miR-125a-5p* correlated with poor prognosis in gastrointestinal cancer patients. Herein, *hsa-miR-125a-5p* was overexpressed in imatinib-treated GIST samples, suggesting that *hsa-miR-125a-5p* functions as a major regulator in the progression of GIST.

KEGG pathway enrichment analysis showed that the *hsa-miR-125a-5p* target genes were mainly enriched in focal adhesion and GnRH signaling pathways. Focal adhesion has been confirmed to play an important role in tumor metastasis [28]. Murata et al. [29] showed that focal adhesion kinase (FAK) was an important mediator, functioning between cells and the extracellular matrix and that FAK localization was related to colorectal carcinogenesis. Wang et

al. [30] reported that mutations of Rho GTPases might affect the progression of gastric cancer through adherent junction and focal adhesion pathway, and a recent study [31] revealed that silencing *profilin-1* could inhibit gastric cancer progression, with the effect most likely mediated by the integrin β 1/FAK pathway. In addition, the GnRH signaling pathway was reported to be involved in tumor pathogenesis and progression in several cancers, such as lung, pancreatic and colon cancer [32].

Further, GnRH agonists have been demonstrated to inhibit ovarian, prostate and breast cancer cell progression [33–35]. GnRH might also significantly decrease angiogenesis in melanoma by regulating the VEGFs pathway [36]. However, the effect of the GnRH signaling pathway in GIST was less reported. Chang et al. [37] identified multiple novel genes associated with metastasis of gastric cancer. Based on our findings, we speculate that *hsa-miR-125a-5p* might play an important role in GIST progression and that the effect of the imatinib-response was likely mediated through the focal adhesion and GnRH signaling pathways. However, the potential mechanism of *hsa-miR-125a-5p* in imatinib resistance in GIST still needs further investigation.

Although *hsa-miR-28-5p* was also over-expressed in drug resistant GIST samples, very little is known about it except that its up-regulation was reported in ovarian cancer tissues compared to adjacent ovarian tissues, and it might promote ovarian tumor growth through down-regulation of N4BP1 *in vivo* [38]. Almeida et al. [7] also revealed

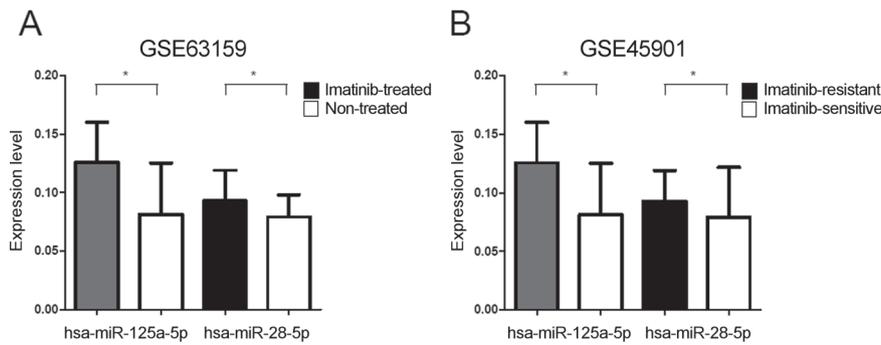


Figure 6. The expression level of *hsa-miR-125a-5p* and *hsa-miR-28-5p* in the GSE63159 (A) and GSE45901 (B) datasets. A) The gray and black columns represent the expression level of *hsa-miR-125a-5p* and *hsa-miR-28-5p* in imatinib-treated GIST samples, respectively and the white columns represent the untreated GIST samples. B) The gray and black columns represent the expression level of *hsa-miR-125a-5p* and *hsa-miR-28-5p* in imatinib resistance GIST samples, respectively, while the white columns represent the imatinib-sensitive GIST samples. * indicates significant difference between the two groups ($p < 0.05$).

that *a-miR-28-5p* suppressed tumor cell proliferation and caused tumor cell apoptosis in colorectal cancer cells.

Although the function of *hsa-miR-28-5p* in GIST progress has not been reported, we speculate that over-expression of *hsa-miR-28-5p* may play an important role in imatinib resistance in GIST patients.

The strengths of this study include the identification of *hsa-miR-28-5p* and *hsa-miR-125a-5p* which provides us with a broader perspective for elucidating the key mechanism associated with imatinib resistance and sensitivity in imatinib-treated GIST patients.

However, there are some limitations that merit further consideration. Firstly, the expression levels of the identified key miRNAs were not verified by performing biological experiments with patient samples. Secondly, our experiments did not exclusively confirm if changing the expression of these two miRNAs can alter the sensitivity to imatinib treatment, and therefore further experiments are required to confirm the findings of this study.

In conclusion, our study based on bioinformatics analysis suggests that the critical miRNAs, such as *hsa-miR-28-5p* and *hsa-miR-125a-5p* are most likely associated with imatinib resistance or sensitivity in imatinib-treated GIST patients. *Hsa-miR-28-5p* and *hsa-miR-125a-5p* may also be considered prognostic markers in imatinib-response in GIST patients.

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