

Differential mRNA and Long Noncoding RNA Expression Profiles in Pediatric B-Cell Acute Lymphoblastic Leukemia Patients

Jing Xia

The Affiliated Wuxi Children's Hospital of Nanjing Medical University

Mengjie Wang

The Affiliated Wuxi Children's Hospital of Nanjing Medical University

Yi Zhu

The Affiliated Wuxi Children's Hospital of Nanjing Medical University

Chaozhi Bu

The Affiliated Wuxi Maternity and Child Health Care Hospital of Nanjing Medical University

Tianyu Li (■ litianyu@njmu.edu.cn)

The Affiliated Wuxi Children's Hospital of Nanjing Medical University

Research Article

Keywords: acute lymphoblastic leukemia, long non-coding RNA, mRNA, next generation sequencing, transcriptome network

Posted Date: October 25th, 2021

DOI: https://doi.org/10.21203/rs.3.rs-879931/v1

License: © ① This work is licensed under a Creative Commons Attribution 4.0 International License. Read Full License

Version of Record: A version of this preprint was published at BMC Pediatrics on January 3rd, 2022. See the published version at https://doi.org/10.1186/s12887-021-03073-5.

Abstract

Background

Long non-coding RNAs (IncRNAs) are transcripts longer than 200 nt that are involved in the pathogenesis and development of various cancers including B cell acute lymphoblastic leukemia (B-ALL). To determine whether IncRNAs are involved in B-ALL, we analyzed the expression profile of IncRNAs and mRNAs in B-ALL.

Methods

We performed RNA sequencing of 10 non-leukemic blood disease donors and 10 B-ALL patients for Gene Ontology (GO) and the Kyoto Encyclopedia of Genes and Genomes (KEGG) analysis. Interactions among mRNAs were predicted using the STRING database. Quantitative real time PCR (qRT-PCR) was performed to verify the RNA-seq data of IncRNAs and mRNAs. Potential functions of subtype-specific IncRNAs were determined by using coexpression-based analysis on distally (trans-pattern) located protein-coding genes.

Results

A total of 1813 differentially expressed genes (DETs) and 2203 IncRNAs were identified. Moreover, 10 dysregulated IncRNAs and 10 mRNAs were randomly selected, and further assessed by RT-qPCR in vitro. Go and KEGG analysis demonstrated that the differentially expressed mRNAs were most closely associated with myeloid leukocyte activation and in transcriptional misregulation in cancer, respectively. In addition, co-expression analysis demonstrated that these IncRNAs, including MSTRG.27994.3, MSTRG.21740.1, ENST00000456341, MSTRG.14224.1 and MSTRG.20153.1, may mediate the pathogenesis and development of B-ALL via IncRNA-mRNA network interactions.

Conclusions

These results showed that several mRNAs and IncRNAs are aberrantly expressed in the bone marrow of B-ALL patients and play potential roles in B-ALL development, and be useful for diagnostic and/or prognostic purposes in pediatric B-ALL.

Background

Acute lymphoblastic leukemia (ALL) is the most common malignancy in children, accounting for approximately 30% of all childhood cancers and approximately 80% of all childhood leukemia. Of these incidences of childhood leukemia, 80% are derived from B-cell progenitors (B-ALL) [1]. B-ALL has one of the highest cure rates of pediatric malignancies [2, 3]. However, although most patients can be cured, higher performance therapies for ALL are required to improve the unfavorable outcomes. Thus, it is necessary to achieve an improved understanding of its potential pathogenesis.

Transcriptome profile analysis is an advanced method that has been applied to reveal the etiology underlying the specific biological processes of ALL. It benefits from ribose nucleic acid sequencing (RNAseq), which is the most extensively used technology for detecting transcriptomes [4]. Previous studies comparing the pathophysiologies of ALL to chronic lymphocytic leukemia (CLL) or comparing the molecular pathogenesis features of adult and pediatric ALL, have focused on analyzing the expression profiles of microRNAs (miRNAs), messenger RNA (mRNA), and miRNA/mRNA networks [5-7]. Long noncoding RNAs (IncRNAs), which are one of the most studied non-coding RNA types, are over 200 nucleotides in length. LncRNAs can act as gene expression modulators at the epigenetic, transcriptional, and post-transcriptional levels. While the roles of IncRNAs in ALL and their underlying mechanisms have been continuously investigated, related data of research articles remain scarce. Fernando et al. demonstrated that the IncRNA CASC15 is fundamental for the cellular survival and proliferation of ALL cells, as it regulates SOX4 expression [8]. Another study has suggested that the IncRNA BALR-6 may have an effect on proliferation and apoptosis in ALL cell lines [9]. In addition, in vitro studies have shown that the lncRNA MALAT1 affects ALL cell proliferation and apoptosis via regulating the miR-205-PTK7 axis. Furthermore, the IncRNA CRNDE has been shown to upregulate CREB expression by suppressing miR-345-5p, promoting cell proliferation, and reducing cell apoptosis in ALL [10, 11]. All of these above studies indicate that IncRNAs could affect the expression of mRNAs via co-expression networks. Nevertheless, the role of IncRNA-mRNA networks in the pathogenesis of B-ALL is not fully understood. This study analyzed the expression profiles of IncRNAs and mRNAs in B-ALL, compared with non-leukemic blood disease controls, by using Next Generation Sequencing (NGS). Specific focus was given to differential mRNA and IncRNA expression levels and IncRNA-mRNA networks to identify the potential underlying mechanism of B-ALL. This approach may provide novel insights into further functional studies, and may reveal new biomarkers for the diagnosis of B-ALL.

Materials And Methods

Patients and samples

Bone marrow samples were obtained at diagnosis from ten pediatric B-ALL patients. Further samples from ten pediatric donors with common blood diseases (CBDs), diagnosed as anemia and agranulocytosis; these were used as controls to lessen the potential variation. All samples were collected form patients at the Affiliated Wuxi Children's Hospital of Nanjing Medical University. The pediatric B-ALL patients included in present study were diagnosed in accordance with the revised French-American-British classification and by analysis of leukemic cells with respect to morphology, immunophenotype, and cytogenetics. Bone marrow samples were submitted to cell separation using Ficoll-Paque (GE Healthcare, USA) and the isolated mononuclear cell fraction was used for the subsequent extraction of the total RNA. The study protocol was approved by the institutional review Board of the Affiliated Wuxi Children's Hospital of Nanjing Medical University (WXCH 2020-08-004). This trail was conducted basing on the written informed consent of all patients, and particularized information is summarized in Table S1.

RNA Extraction, strand-specific library construction and sequencing

Total RNA from mononuclear cells was extracted using a Trizol reagent kit (Invitrogen, Carlsbad, CA, USA) consistent with the manufacturer's instructions. RNA quality was appraised using an Agilent 2100 Bioanalyzer (Agilent Technologies, Palo Alto, CA, USA). After total RNA was extracted, rRNAs were removed to retain mRNAs and non-coding RNAs (ncRNAs). The enriched mRNAs and IncRNAs were fragmented into short pieces using a The fragmentation buffer was used to fragment into short, and were then reverse transcribed into complementary deoxyribose nucleic acid (cDNA) using random primers. Second-strand cDNA were synthesized using DNA polymerase I, RNase H, dNTP (dUTP instead of dTTP), and buffer. Next, the cDNA fragments were purified using a QiaQuick polymerase chain reaction (PCR) extraction kit (Qiagen, Venlo, The Netherlands). They were then end repaired, poly (A) added, and bound to Illumina sequencing adapters. Then, the digested second-strand cDNA using Uracil-N-Glycosylase (UNG) were size selected by programming steps including agarose gel electrophoresis, PCR amplified, and sequenced using Illumina HiSeqTM 4000 (or other platforms) by Gene Denovo Biotechnology Co. (Guangzhou, China).

Filtering of Clean Reads

Reads obtained from the sequencing machines included raw reads containing adapters or, which could have affected the following assembly and analysis. To procure high quality clean reads, raw reads including low quality bases or adapters that may interfere subsequent analysis were filtered using fastp (version 0.18.0). The parameters were as follows: 1) removing reads containing adapters, 2) removing reads containing more than 10% of unknown nucleotides (N), and 3) removing low quality reads containing more than 50% of low quality (Q-value≤20) bases.

Differentially expressed transcripts (DETs) Analysis

The DETs of coding RNAs and IncRNAs were analyzed respectively. DESeq2 software was used to analyze differential expression of mRNAs and IncRNAs between two different groups (CBD group versus B-ALL group). Those transcripts with the parameters of a false discovery rate (FDR) of < 0.05 and an absolute fold change (FC) of \geq 2 were considered to be differentially expressed. Enrichment analysis of Gene Ontology (GO) functions, Kyoto Encyclopedia of Genes (KEGG) pathways, Disease Ontology (DO), Reactome Enrichment Analysis, and Gene Set Enrichment Analysis (GESA) were performed by choosing the differentially expressed coding RNAs.

Functional Enrichment Analysis

GO enrichment analysis was performed basing on three terms including biological processes, cellular components and molecular functions (http://www.geneontology.org/). KEGG enrichment analysis distinguished significantly enriched biological pathways. DO enrichment analysis identified significantly enriched human disease DO terms in DETs. Reactome enrichment analysis identified significantly enriched reactions including signaling, innate and acquired immune function, transcriptional regulation,

translation, apoptosis, and classical intermediary metabolism in DETs. GESA analysis was performed using the software packages GSEA and MSigDB to identify genes in specific GO terms\pathways\DO terms showing significant differences between the two groups. All above analyses meeting the condition of p < 0.05 were considered to be significant.

LncRNA-mRNA association analysis

The IncRNA-mRNA association analysis was based on IncRNA trans-regulation analysis to reveal the interaction between IncRNA and mRNA and identify target genes of IncRNAs.

Validation of RNA-seq results

A number of mRNAs and IncRNAs between the paired groups were selected to validate the RNA-seq data by quantitative real time PCR (qRT-PCR). After determining the best annealing temperatures, qRT-PCR was performed on the ABI 7500 system (Applied Biosystems, Foster City, CA,USA) to detect the relative expression levels of mRNA and IncRNA using SYBR Premix ExTaq II (DRR820; TaKaRa). The relative expression levels of each mRNA/IncRNA were calculated using comparative cycle threshold method ($\Delta\Delta$ Ct). Standardization was performed based on the internal reference of GAPDH. All data are presented herein as the average of three independent experiments.

Statistical analysis

Hypergeometric Test was used to calculate the enrichment of GO terms, KEGG pathways, and DO enrichment analysis to show the significance of enrolled transcripts. Quantitative data was assessed with the use of two independent samples t-test. Data are presented herein as the mean \pm standard deviation. The edgeR software package (version 3.28.1) was used to identify differentially expressed lncRNAs and mRNAs between paired groups. p < 0.05 was considered to indicate a statically significant difference.

Results

Differentially expressed IncRNAs and mRNAs in B-ALL

Coding Potential Calculator (CPC) 2 and Coding-Non-Coding Index (CNCI), two efficient tools for distinguish protein-coding and non-coding sequences, were used to perform prediction analysis of the coding capabilities of novel IncRNAs. The number of non-coding transcripts were 8049 and 7289 following CPC2 and CNCI analysis, respectively (Fig. S1A). The overlaps of novel 6283 non-coding IncRNAs were enrolled for subsequent expression profile analysis combined with the known IncRNAs. According to the position relation of IncRNAs and adjacen mRNAs, IncRNAs were classified into six categories: 47.0% were intergenic, 11.2% were antisense, 32.6% were sense, 3.0% were intronic, 4.4% were bidirectional, and 1.8 % were others (Fig. S1B).

Histogram, volcano plots and heat maps were used for assessing gene expression variation between CBD and B-ALL groups. In total, 2203 IncRNAs displayed differential expression in B-ALL, including 1592 upregulated IncRNAs and 611 downregulated IncRNAs (Fig. 1A). Of 1813 mRNAs that showed differential

expression, 1076 were upregulated and 737 were downregulated (Fig. 1B). Among them, 376 IncRNAs and 105 mRNAs were significantly upregulated, and 15 IncRNAs and 4 mRNAs were significantly downregulated >5(\log_2 FC) in B-ALL. In-depth analysis showed that 16 IncRNAs and 2 mRNAs (IGKV10R2-108 and IRX2) were significantly upregulated >10 (\log_2 FC) in B-ALL. Hierarchical clustering analysis showed systematic variations in the expression of IncRNAs and mRNAs among samples. The data suggested that the expression of IncRNAs and mRNAs in B-ALL differ from those in CBD group. All differentially expressed IncRNAs are listed in Supplementary Table S2 and differentially mRNAs expressed are listed in Supplementary Table S3.

Validation of differentially expressed IncRNAs and differentially expressed mRNAs using qPCR

Based on relatively high variation, FC>2, and *p*<0.05, we selected 10 lncRNAs and 10 mRNAs to assess the RNA-seq results by qRT-PCR using PCR primers as documented in Table S4. The present results indicated that the lncRNA expression levels of ENST00000473898, ENST00000649077, ENST00000421323, ENST00000662821 and MSTRG.13057.1 were upregulated (Fig. 2A), whereas the expression levels of ENST00000553415, ENST00000643525, MSTRG.33538.27, ENST00000650778 and ENST00000662196 were downregulated (Fig. 2B). Moreover, the mRNA expression levels of ENSG00000170561 (IRX2), ENSG00000135333 (EPHA7), ENSG00000213934 (HBG1), ENSG00000170549 (IRX1) and ENSG00000102755 (FLT1) were upregulated (Fig. 2C), while ENSG00000169429 (CXCL8), ENSG00000123689 (GOS2), ENSG00000163421 (PROK2), ENSG00000163739 (CXCL1), ENSG00000112715 (VEGFA) were downregulated (Fig. 2D). The function of these genes was associated with metabolism, cell processing and inflammatory response. Therefore, the qPCR results were conclusively coincided with those identified in RNA-seq data.

Enrichment analysis of differentially expressed genes

The GO and KEGG pathway analyses of differentially expressed mRNAs were performed to provide a clue about the B-ALL disease process, which may also reveal the role of IncRNAs in B-ALL. We enrolled all differentially expressed mRNAs for the GO analysis and found that the most enriched GO terms by upregulated and downregulated transcripts were most highly involved in myeloid leukocyte activation, plasma membrane and calcium ion binding in corresponding biological process category (Fig. 3A), cellular component category (Fig. 3B) and molecular function category, respectively (Fig. 3C). KEGG pathway analysis results demonstrated that the differentially expressed mRNAs were mainly enriched in transcriptional misregulation in cancer, leishmaniasis, amoebiasis, herpes simplex infection and Rap1 signaling pathway (Fig. 3D).

Disease Ontology (DO) analysis demonstrated that highly enriched disease were lymphoblastic leukemia, arthritis and bone inflammation disease, which were associated with the etiology or clinical symptom of B-ALL (Fig. S2A). In addition, the Reactome considered as an improved pathway database was used for Entities (nucleic acids, proteins, complexes and small molecules) annotation of enriched pathways.

Results showed that differentially expressed mRNAs were involved in neutrophil DETranulation, innate immune system and exocytosis of specific granule lumen proteins in B-ALL (Fig. S2B).

Functional analyses of the mRNAs

To further explore the function of genes at the protein level, and to reveal the core mRNAs in the cellular process of B-ALL, Search Tool for the Retrieval of Interacting Genes/Proteins (STRING) was used to filter functional genes constructing an interaction network for annotating structural and functional properties of proteins[12]. Considering that too many protein showing interactive function, the threshold of confidence score >0.9 and FDR<0.01 were adopted to assess the interactive strength of protein-protein encoded by the DETs. The network contains 1191 nodes and 2014 edges with protein-protein interaction enrichment p-value of 1.0e-16 (Fig. 4A). The highest scores were 0.99 among which protein pairs of VEGFA and FLT1 drew our attention in Rap1 signaling pathway that was one of top 5 enriched pathway by KEGG analysis due to the two encoding genes coincidently differentially expressed in Rap1 signaling pathway. It was demonstrated that whether Rap1 activation was involved in the notch-dependent leukemogenicity of T-ALL [13, 14]. The networks enriched in Rap1 signaling pathway were performed with STRING (confidence score>0.9, FDR<0.01) to illustrate the interaction of proteins in this pathway (Fig. 4B). Although another enriched pathway of Glycolysis/Gluconeogenesis did not conform to the threshold of FDR<0.01, DETs including ENO1, a key enzyme functioning in glycolysis process were enriched in Glycolysis/Gluconeogenesis pathway following the GSEA analysis (NES=-1.553, p=0.005, FDR=0.036) (Fig. 4C), which indicated that the pathogenesis of B-ALL were associated with the deregulated expression of glycometabolism-related genes resulting in abnormal energy supply.

LncRNA/mRNA co-expression network and GO/KEGG annotations in B-ALL

It is well known that antisense IncRNAs may regulate the gene silencing, transcription and stability of mRNA by binding to its sense strand, and IncRNAs could modulate the expression of neighboring coding genes by cis-pattern [15, 16]. Trans-pattern analysis independent on the position of the encoding gene is able to construct correlation or co-expression network between IncRNA and mRNA. We performed transpattern analysis by enrolling the differently expressed IncRNA and mRNA (FC≥2, FDR<0.01) (Table S5), and found that basing on the threshold of Pearson's correlation (>0.99), IncRNAs MSTRG.27994.3, MSTRG.21740.1, ENST00000456341, MSTRG.14224.1 and MSTRG.20153.1 seemed to be the vital parts in the network interacting with at least 7 DETs that contained a differently expressed mRNA of PRKCZ that acts in the Rap1 signaling pathway (Fig. 5), among which MSTRG.21740.1 interacted with 8 mRNAs. Therefore, the present results suggested that these IncRNAs may serve important roles in the pathogenesis of B-ALL.

In addition, the enriched GO term was used to examine the potential function of mRNAs related to differentially expressed IncRNAs following trans-pattern analysis. For biological process, the top 3 enriched term were biological adhesion (GO:0022610), homophilic cell adhesion via plasma membrane

adhesion molecules (GO:0007156), and cell-cell adhesion via plasma-membrane adhesion molecules (GO:00398742) (Fig. S3A); for cellular component, the top 3 enriched term were plasma membrane part (GO:0044459), cell periphery (GO:0071944), and intrinsic component of plasma membrane (GO:0031226) (Fig. S3B); for molecular function, the top 3 enriched term were calcium ion binding (GO:0005099), transmembrane receptor protein tyrosine kinase activity (GO:0004714), and transmembrane receptor protein kinase activity (GO:0019199) (Fig. S3C). In this study, the majority of mRNAs appear to be associated with immune response, metabolic process, and transcriptional activation. Furthermore, the KEGG results showed that lncRNA/mRNA co-expression analysis by trans-pattern were mainly involved in axon guidance, herpes simplex infection, Rap1 signaling pathway, transcriptional misregulation in cancer and MicroRNAs in cancer (Fig. S3D), which suggested that Rap1 signaling pathway also played an important role in the lncRNA/mRNA network.

Discussion

Previous studies regarding the transcriptic results of B-ALL patients have identified a number of lncRNAs and vital protein—coding genes that are associated with the etiology or clinical outcome of B-ALL [17–19]. To the best of the authors' knowledge, although some studies have performed transcriptome profile analysis to investigate the role of lncRNA in the pathogenesis of B-ALL, the small sample sizes of parallel their controls and the restricted numbers of isolated cell types limited the general conclusions of these studies [20–22]. Moreover, microarray data prevented the identification of novel transcripts. In the present study, a comprehensive analysis of lncRNA and mRNA profiling data was conducted between B-ALL patients and CBD controls by NGS. Several differently expressed lncRNAs and mRNAs were identified, in which some were validated by qRT-PCR; functional annotations were also carried out. Furthermore, lncRNA-mRNA network analyses were preformed to extract candidate lncRNAs interacting with the aberrantly expressed mRNAs, revealing that five key lncRNAs were co–expressed with at least seven mRNAs in the network, bases on a very high threshold of 0.99 for the Pearson's correlation. These findings could promote the understanding of the functional mechanism through which lncRNA interacts with mRNA, thus playing an important role in B-ALL pathogenesis.

The biological functions and potential pathways of mRNAs associated with differentially expressed lncRNAs were preliminarily predicted by GO and KEGG pathway analyses. Extraordinarily, significantly enriched pathways were closely associated with signal transduction, metabolic process, infections, and the immune system, which all play important roles in B-ALL development and disease treatment. The KEGG and co-expression analysis results showed that the Rap1 signaling pathway was exclusively highly enriched in B-ALL, when compared to controls. It was demonstrated that the Rap1 signaling pathway is associated with leukemia cell adhesion and migration, suggesting that Rap1 could participate the process of notch activation and leukemogenicity of T-ALL. This suggests that the Rap1 signal may function in the notch-dependent B-ALL development and its progression[14, 23, 24]. Infante et al. reported that Rap1b is a subtype of Rap1, and that its depletion could be used to reduce tissue invasion in T-ALL. Furthermore, another integrative network analysis of pediatric acute lymphoblastic leukemia revealed that gene expression and methylation consistently targeted the Rap1 signaling pathway [25]. In addition, the

direct activation of Rap1 can lessen the cellular actions of leptin and correct glucose imbalance in obese mouse, which suggests that Rap1 is associated with energy metabolism [26]. VEGFA and FLT1, two differentially expressed genes, presented the highest correlation score of 0.99 by STRING in the Rap1 signaling pathway. Previous studies have shown that increased levels of Helios Treg cells promoted angiogenesis in the bone marrow of ALL mice via the VEGFA/VEGF receptor 2 pathway [27]. Furthermore, FLT-1 is generally expressed in pediatric ALL and VEGFA/FLT-1 signaling enhances the migration and survival of leukemia [28, 29]. FLT-1 activation on ALL cells results in cell migration and proliferation in vitro. In addition, FLT-1 neutralization affects leukemia localization, increases leukemia apoptosis, and impedes the exit of ALL cells, thus prolonging the survival of inoculated mice. All of the above results indicate that the VEGFA/FLT-1 interaction could play important roles in regulating the development of B-ALL. PRKCZ is a calcium- and diacylglycerol-independent serine/threonine protein kinase that functions in the phosphatidylinositol 3-kinase (PI3K) pathway and the mitogen-activated protein (MAP) kinase cascade. It is also involved in NF-kappa-B activation, mitogenic signaling, cell proliferation, and inflammatory response. AKT, PI3K2CB, and some other genes also feature in the pathway upstream of PRKCZ. The PI3K/Akt and NF-kB signaling pathways have been extensively demonstrated to participate in the regulation of migration, and in the viability of leukemic cells, both in vivo and in vitro [30–32]. It has also been reported that the knockdown of PRKCZ leads to more rapid losses of DNA mismatch repair enzyme (MSH2) proteins, resulting in significant reductions in DNA mismatch repair (MMR) and increased resistance to thiopurines for ALL [33]. Arsenic trioxide (ATO) has been demonstrated to interrupt the function of the PI3K/Akt pathway in ALL; PRKCZ may be responsible for the provocation of resistance to ATO [34]. The most aberrant upregulated IncRNA, ENST00000473898, showed a high correlation score of 0.95 with PRKCZ, suggesting that an in-depth study should be conducted to elucidate its potential regulatory mechanism.

The IncRNA-mRNA network constructed in the present study revealed that some IncRNAs may exert significant effects in B-ALL by regulating downstream mRNAs. Affinito et al. specially constructed a IncRNAs-mRNAs co-expression network in childhood B-ALL, based on previous RNA-sequencing experiments [35]. In said experiments, gene expression profiling using RNA-seq of leukemic cells purified from three pediatric B-ALL patients (compared with that observed in mature B cells from the peripheral blood of three healthy donors) highlighted that 24 key IncRNAs and their co-expressed mRNAs may play important roles in B-ALL pathogenesis [36]. Unlike this study, however, here larger sample sizes of B-ALL patients were used alongside quantity-matched controls. Furthermore, the control group used the same cell type of mononuclear cells derived from bone marrow. In addition, trans-patterns were used to predict the co-expression network of IncRNAs with mRNAs, unlike the guilt-by-association approach of the aforementioned previous study. The present study also obtained more high thresholds of FDR (<0.01) and high Pearson correlation scores (0.99). Here, five novel IncRNAs were identified, which interacted with at least seven mRNAs, including FAT1, SOX11, and EYA4. The functional roles of these IncRNAs remain unknown. Nevertheless, according to the alternative mRNAs in coexpression networks, these lncRNAs may exert exclusive characters by regulating downstream mRNAs. Among them, FAT1 was highly expressed in a large proportion of cases of T-ALL and B-ALL was implicated in Wnt signaling and hippo

signaling. FAT1 was also found to cooperate with NOTCH in driving T-ALL *in vivo*; this suggests that it might be a potential biomarker in carcinogenic roles [37, 38]. Moreover, previous studies have shown SOX11 to be a useful diagnostic marker for mantle cell lymphoma [39]. It has been reported that the high expression of SOX11 leads to alterations of gene expression that are typically associated with cell adhesion, migration, and differentiation. Furthermore, its expression marks a group of patients with good outcomes [40]. Furthermore, Huang et al. demonstrated that AML1-ETO-fused protein triggers the epigenetic silencing of the EYA4 gene, contributing to leukemogenesis in t(8;21) AML. This suggests that the EYA4 gene might be a novel therapeutic target for AML. In addition, although here the co–expression lncRNA-mRNA pairs were captured from RNA-Seq data, it is well established that co–expression correlation does not imply causation. Thus, gene perturbation experimental data would be necessary to gain insight into the possible regulatory relationships, or to infer causal gene co–expression patterns.

Conclusions

In conclusion, the present study identified differentially expressed IncRNAs and mRNAs between CBD and B-ALL, via NGS. Some of these genes were randomly selected and assessed by qRT-PCR *in vitro*. Go and KEGG analyses were performed, and the obtained co-expression network indicated that the IncRNAs MSTRG.27994.3, MSTRG.21740.1, ENST00000456341, MSTRG.14224.1, and MSTRG.20153.1 may have bioactive effects in B-ALL. Therefore, the present study may provide pathological sagacity into the mechanisms underlying B-ALL. More importantly, these results provide a basis for further studies into the function and mechanism of special mRNAs and IncRNAs in B-ALL. However, further functional investigations are required to confirm these results.

Abbreviations

CLL: chronic lymphocytic leukemia

B-ALL: B cell acute lymphoblastic leukemia

IncRNA: long non-coding RNA

CBD: common blood diseases

DETs: Differentially expressed transcripts

GO: Gene Ontology

KEGG: Kyoto Encyclopedia of Genes

FC: fold change

Declarations

Availability of data and materials

All data analyzed during this study are included in this article.

Acknowledgements: We are grateful to thank Guangzhou Genedenovo Biotechnology Co., Ltd for assisting in sequencing and bioinformatics analysis.

Funding: This research was funded by Key Program of Jiangsu Health Commission (ZDA2020013), Wuxi Medical Development Discipline (FZXK001) and Science and Technology Development Fund Project of Nanjing Medical University (NMUB2020285).

Author information

Affiliations

Department of Pediatric Laboratory, The Affiliated Wuxi Children's Hospital of Nanjing Medical University, Wuxi 214000, Jiangsu, China.

Department of hematology & oncology, The Affiliated Wuxi Children's Hospital of Nanjing Medical University, Wuxi 214000, Jiangsu, China.

Center of Reproductive Medicine, State Key Laboratory of Reproductive Medicine, Research Institute for Reproductive Health and Genetic Diseases, The Affiliated Wuxi Matemity and Child Health Care Hospital of Nanjing Medical University, Wuxi 214002, China.

Jing Xia, Mengjie Wang, Yi Zhu, Chaozhi Bu & Tianyu Li

Contributions

JX, who is first author, conducted the work including the data collection and data analysis, and original draft preparation. As the co-first author, MW performed patient recruitment and statistical analyses. YZ performed the extraction of total RNA of the bone marrow samples. CB and TL performed the conceptualization and supervision for this study, and reviewed and edited the manuscript.

Corresponding author

Correspondence to Chaozhi Bu & Tianyu Li.

Ethics declarations

Ethics approval and consent to participate

Not applicable.

Conflict of interest

The authors declare no conflict of interest.

Consent for publication

Not applicable.

References

- 1. Pui CH. Recent research advances in childhood acute lymphoblastic leukemia. J Formos Med Assoc. 2010;109:777-87.
- 2. Pui CH, Robison LL, Look AT. Acute lymphoblastic leukaemia. Lancet. 2008;371:1030-43.
- 3. Hunger SP, Lu X, Devidas M, Camitta BM, Gaynon PS, Winick NJ, et al. Improved survival for children and adolescents with acute lymphoblastic leukemia between 1990 and 2005: a report from the children's oncology group. J Clin Oncol. 2012;30:1663–9.
- 4. Wang Z, Gerstein M, Snyder M. RNA-Seq: a revolutionary tool for transcriptomics. Nat Rev Genet. 2009;10:57–63.
- 5. Zhang S, Zhang Q, Yin J, Wu X. Overlapped differentially expressed genes between acute lymphoblastic leukemia and chronic lymphocytic leukemia revealed potential key genes and pathways involved in leukemia. J Cell Biochem. 2019;120:15980–8.
- 6. Almeida RS, Costa ESM, Coutinho LL, Garcia Gomes R, Pedrosa F, Massaro JD, et al. MicroRNA expression profiles discriminate childhood T- from B-acute lymphoblastic leukemia. Hematol Oncol. 2019;37:103–12.
- 7. Chen X, Guo Z, Fan S, Sun L, Li H, Zhou J, et al. Integrating microRNA and mRNA expression in rapamycin-treated T-cell acute lymphoblastic leukemia. Pathol Res Pr. 2019;215:152494.
- 8. Fernando TR, Contreras JR, Zampini M, Rodriguez-Malave NI, Alberti MO, Anguiano J, et al. The IncRNA CASC15 regulates SOX4 expression in RUNX1-rearranged acute leukemia. Mol Cancer. 2017;16:126.
- 9. Rodríguez-Malavé NI, Fernando TR, Patel PC, Contreras JR, Palanichamy JK, Tran TM, et al. BALR-6 regulates cell growth and cell survival in B-lymphoblastic leukemia. Mol Cancer. 2015;14:214.
- 10. Song Y, Guo NH, Zheng JF. LncRNA-MALAT1 regulates proliferation and apoptosis of acute lymphoblastic leukemia cells via miR-205-PTK7 pathway. Pathol Int. 2020;70:724–32.
- 11. Wang W, Wu F, Ma P, Gan S, Li X, Chen L, et al. LncRNA CRNDE Promotes the Progression of B-cell Precursor Acute Lymphoblastic Leukemia by Targeting the miR-345-5p/CREB Axis. Mol Cells. 2020;43:718–27.
- 12. Szklarczyk D, Franceschini A, Wyder S, Forslund K, Heller D, Huerta-Cepas J, et al. STRING v10: protein-protein interaction networks, integrated over the tree of life. Nucleic Acids Res. 2015;43 Database issue:D447-52.
- 13. Petri A, Dybkær K, Bøgsted M, Thrue CA, Hagedorn PH, Schmitz A, et al. Long Noncoding RNA Expression during Human B-Cell Development. PLoS One. 2015;10:e0138236.

- 14. Infante E, Heasman SJ, Ridley AJ. Statins inhibit T-acute lymphoblastic leukemia cell adhesion and migration through Rap1b. J Leukoc Biol. 2011;89:577–86.
- 15. Gil N, Ulitsky I. Regulation of gene expression by cis-acting long non-coding RNAs. 2020;21:102-17.
- 16. Alkan C, Karakoç E, Nadeau JH, Sahinalp SC, Zhang K. RNA-RNA interaction prediction and antisense RNA target search. J Comput Biol. 2006;13:267–82.
- 17. James AR, Schroeder MP, Neumann M, Bastian L, Eckert C, Gökbuget N, et al. Long non-coding RNAs defining major subtypes of B cell precursor acute lymphoblastic leukemia. J Hematol Oncol. 2019;12:8.
- 18. Rosa R, Villegas-Ruíz V, Caballero-Palacios MC, Pérez-López El, Murata C, Zapata-Tarres M, et al. Expression of ZNF695 Transcript Variants in Childhood B-Cell Acute Lymphoblastic Leukemia. Genes (Basel). 2019;10.
- 19. Fernando TR, Rodriguez-Malave NI, Waters E V, Yan W, Casero D, Basso G, et al. LncRNA Expression Discriminates Karyotype and Predicts Survival in B-Lymphoblastic Leukemia. Mol Cancer Res. 2015;13:839–51.
- 20. Lajoie M, Drouin S, Caron M, St-Onge P, Ouimet M, Gioia R, et al. Specific expression of novel long non-coding RNAs in high-hyperdiploid childhood acute lymphoblastic leukemia. PLoS One. 2017;12:e0174124.
- 21. Cuadros M, Andrades Á, Coira IF, Baliñas C, Rodríguez MI, Álvarez-Pérez JC, et al. Expression of the long non-coding RNA TCL6 is associated with clinical outcome in pediatric B-cell acute lymphoblastic leukemia. Blood Cancer J. 2019;9:93.
- 22. Bárcenas-López DA, Núñez-Enríquez JC, Hidalgo-Miranda A, Beltrán-Anaya FO, May-Hau DI, Jiménez-Hernández E, et al. Transcriptome Analysis Identifies LINC00152 as a Biomarker of Early Relapse and Mortality in Acute Lymphoblastic Leukemia. Genes (Basel). 2020;11.
- 23. Doi K, Imai T, Kressler C, Yagita H, Agata Y, Vooijs M, et al. Crucial role of the Rap G protein signal in Notch activation and leukemogenicity of T-cell acute lymphoblastic leukemia. Sci Rep. 2015;5:7978.
- 24. Wang SF, Aoki M, Nakashima Y, Shinozuka Y, Tanaka H, Taniwaki M, et al. Development of Notch-dependent T-cell leukemia by deregulated Rap1 signaling. Blood. 2008;111:2878–86.
- 25. Sanchez R, Mackenzie SA. Integrative Network Analysis of Differentially Methylated and Expressed Genes for Biomarker Identification in Leukemia. Sci Rep. 2020;10:2123.
- 26. Kaneko K, Lin HY, Fu Y, Saha PK, De la Puente-Gomez AB, Xu Y, et al. Rap1 in the VMH regulates glucose homeostasis. JCl Insight. 2021;6.
- 27. Li X, Li D, Shi Q, Huang X, Ju X. Umbilical cord blood-derived Helios-positive regulatory T cells promote angiogenesis in acute lymphoblastic leukemia in mice via CCL22 and the VEGFA-VEGFR2 pathway. Mol Med Rep. 2019;19:4195–204.
- 28. Diffner E, Gauffin F, Anagnostaki L, Nordgren A, Gustafsson B, Sander B, et al. Expression of VEGF and VEGF receptors in childhood precursor B-cell acute lymphoblastic leukemia evaluated by immunohistochemistry. J Pediatr Hematol Oncol. 2009;31:696–701.

- 29. Xiu B, Zhang W, Huang B, Chen J, Lu H, Fu J, et al. Genetic inhibition of vascular endothelial growth factor receptor-1 significantly inhibits the migration and proliferation of leukemia cells and increases their sensitivity to chemotherapy. Oncol Rep. 2013;29:2030–8.
- 30. Shi YF, Huang ZY, Huang YS, Dong RJ, Xing CY, Yu K, et al. Interaction between CD9 and PI3K-p85 activates the PI3K/AKT signaling pathway in B-lineage acute lymphoblastic leukemia. Oncol Rep. 2021;46.
- 31. Richter A, Fischer E, Holz C, Schulze J, Lange S, Sekora A, et al. Combined Application of Pan-AKT Inhibitor MK-2206 and BCL-2 Antagonist Venetoclax in B-Cell Precursor Acute Lymphoblastic Leukemia. Int J Mol Sci. 2021;22.
- 32. Lin C, Chen D, Xiao T, Lin D, Lin D, Lin L, et al. DNA methylation-mediated silencing of microRNA-204 enhances T cell acute lymphoblastic leukemia by up-regulating MMP-2 and MMP-9 via NF-κB. 2021;25:2365–76.
- 33. Diouf B, Cheng Q, Krynetskaia NF, Yang W, Cheok M, Pei D, et al. Somatic deletions of genes regulating MSH2 protein stability cause DNA mismatch repair deficiency and drug resistance in human leukemia cells. Nat Med. 2011;17:1298–303.
- 34. Roszak J, Smok-Pieniążek A, Nocuń M, Stępnik M. Characterization of arsenic trioxide resistant clones derived from Jurkat leukemia T cell line: focus on PI3K/Akt signaling pathway. Chem Biol Interact. 2013;205:198–211.
- 35. Smaldone G, Beneduce G, Incoronato M, Pane K, Franzese M, Coppola L, et al. KCTD15 is overexpressed in human childhood B-cell acute lymphoid leukemia. Sci Rep. 2019;9:20108.
- 36. Affinito O, Pane K, Smaldone G, Orlandella FM, Mirabelli P, Beneduce G, et al. IncRNAs-mRNAs Co-Expression Network Underlying Childhood B-Cell Acute Lymphoblastic Leukaemia: A Pilot Study. Cancers (Basel). 2020;12.
- 37. de Bock CE, Ardjmand A, Molloy TJ, Bone SM, Johnstone D, Campbell DM, et al. The Fat1 cadherin is overexpressed and an independent prognostic factor for survival in paired diagnosis-relapse samples of precursor B-cell acute lymphoblastic leukemia. Leukemia. 2012;26:918–26.
- 38. de Bock CE, Down M, Baidya K, Sweron B, Boyd AW, Fiers M, et al. T-cell acute lymphoblastic leukemias express a unique truncated FAT1 isoform that cooperates with NOTCH1 in leukemia development. Haematologica. 2019;104:e204–7.
- 39. Lee W, Shin E, Kim BH, Kim H. Diagnostic accuracy of SOX11 immunohistochemistry in mantle cell lymphoma: A meta-analysis. PLoS One. 2019;14:e0225096.
- 40. Grönroos T, Mäkinen A, Laukkanen S, Mehtonen J, Nikkilä A, Oksa L, et al. Clinicopathological features and prognostic value of SOX11 in childhood acute lymphoblastic leukemia. Sci Rep. 2020;10:2043.

Figures

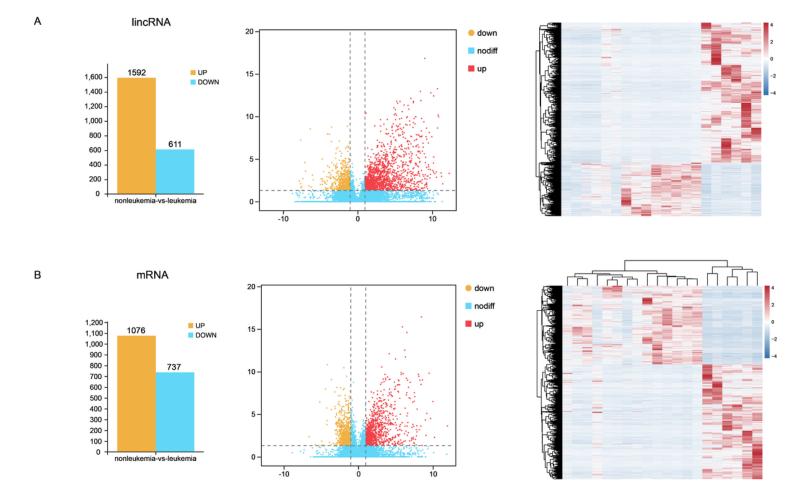


Figure 1

Histogram, volcano plots and heat map showing expression profiles of IncRNAs. (A) and mRNAs (B) in B-ALL. Left panels, histogram visualizing the differentially expressed IncRNAs and mRNAs. Middle and right panels, volcano plots and heat maps showing significantly changed IncRNAs and mRNAs with fold change≥2.0, respectively (p<0.05; FDR<0.05).

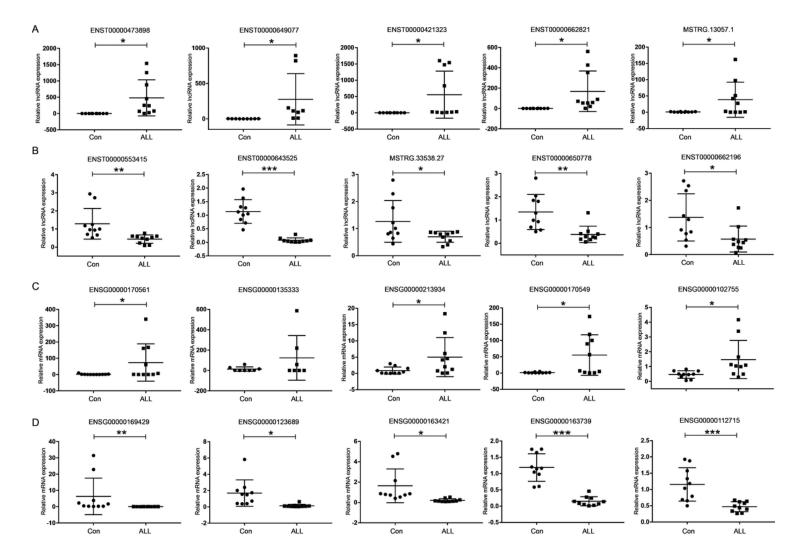


Figure 2

The dysregulated lncRNAs and mRNAs by qPCR in vitro. (A) Validation of upregulated lncRNAs; (B) Validation of downregulated lncRNAs; (C) Validation of upregulated mRNAs; (D) Validation of downregulated mRNAs. *p < 0.05, **p < 0.01, and ***p < 0.001.

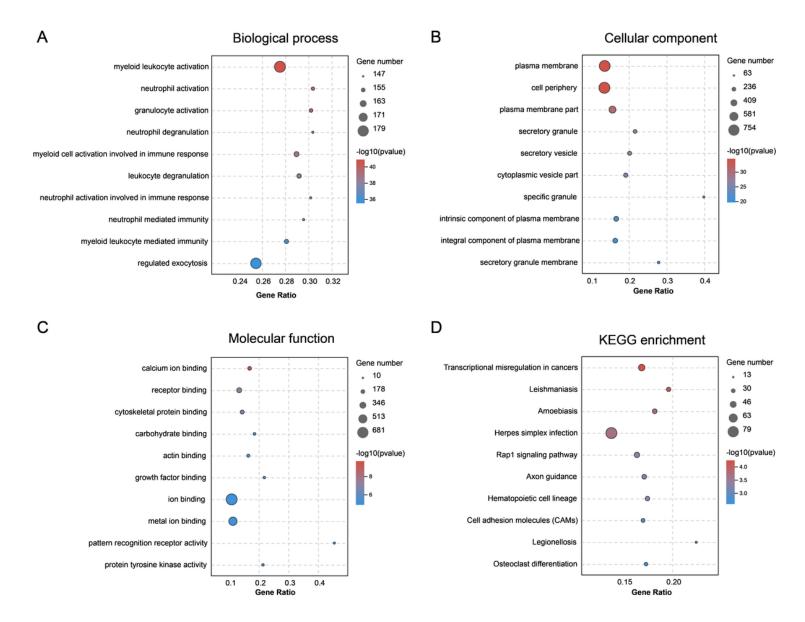


Figure 3

Gene ontology and pathway analysis predicting the potential function of the differentially expressed mRNAs with top 10 enrichment scores in B-ALL. (A) Biological process category. (B) Cellular component category. (C) Molecular function category. (d) Canonical signaling pathways.

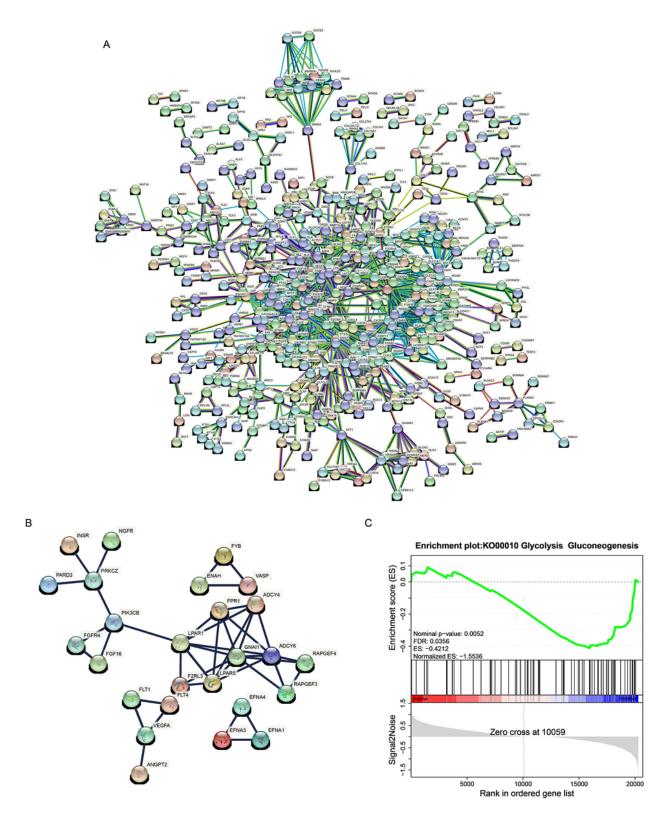


Figure 4

Protein-protein interaction networks by STRING and GSEA analysis of enriched pathway. (A) STRING was used to construct the DETs ($FC \ge 2$, p<0 .05, FDR<0.01) network and confidence score was set as the highest (>0.9). (B) Subnetwork of all DETs enriched in Rap1 signaling pathway. (C) GSEA analysis of Starch and sucrose metabolism pathway.

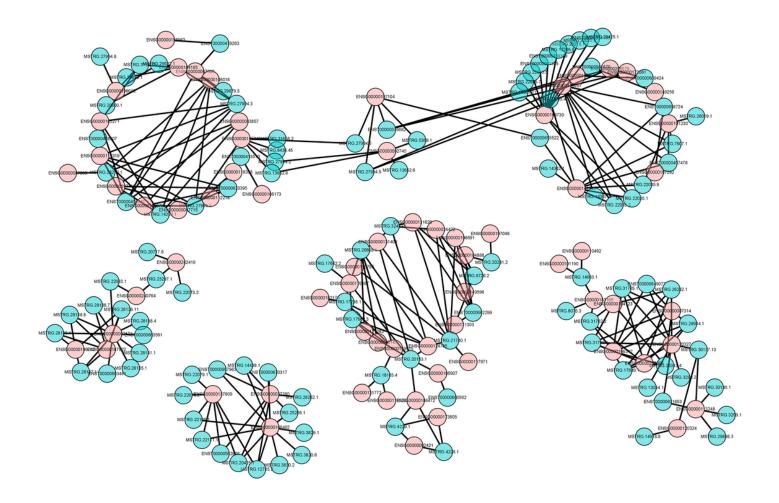


Figure 5

Co-expression networks constructed basing on the Pearson's correlation (>0.99). Pink, mRNA; pale green, lncRNA.

Supplementary Files

This is a list of supplementary files associated with this preprint. Click to download.

- SupplementalFigureS1S3.docx
- SupplementaltableS1S5.xlsx