Small RNA-Seq Reveals Similar miRNA Transcriptome in Children and Young Adults with T-ALL and Indicates miR-143-3p as Novel Candidate Tumor Suppressor in This Leukemia

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Abstract: We aimed to identify miRNAs and pathways specifically deregulated in adolescent and young adult (AYA) T-ALL patients. Small RNA-seq showed no major differences between AYA and pediatric T-ALL, but it revealed downregulation of miR-143-3p in T-ALL patients. Prediction algorithms identified several known and putative oncoproteins targeted by this miRNA, including KRAS, FGF1, and FGF9. Pathway analysis indicated signaling pathways related to cell growth and proliferation, including FGFR signaling and PI3K-AKT signaling, with the majority of genes overrepresented in these pathways being predicted targets of hsa-miR-143-3p. By luciferase reporter assays, we validated direct interactions of this miRNA with KRAS, FGF1, and FGF9. In cell proliferation assays, we showed reduction of cell growth upon miR-143-3p overexpression in two T-ALL cell lines. Our study is the first description of the miRNA transcriptome in AYA T-ALL patients and the first report on tumor suppressor potential of miR-143-3p in T-ALL. Downregulation of this miRNA in T-ALL patients might contribute to enhanced growth and viability of leukemic cells. We also discuss the potential role of miR-143-3p in FGFR signaling. Although this requires more extensive validation, it might be an interesting direction, since FGFR inhibition proved promising in preclinical studies in various cancers.

Keywords: T-cell acute lymphoblastic leukemia (T-ALL); adolescents and young adults (AYA); miR-143-3p; tumor suppressor miRNA; targeting FGFR signaling; miRNA-seq; target prediction; pathway analysis; personalized medicine

1. Introduction

T-cell acute lymphoblastic leukemia is an aggressive hematological malignancy arising from T-cell precursors, diagnosed in all age groups. The group of adolescents and
young adults (AYA) is most frequently defined by 15–39 years of age at diagnosis [1]. The transitional age of AYA patients means that younger patients (<18 years of age) are treated in pediatric centers according to pediatric-oriented protocols, while those over the age of 18 are treated in the centers for adults with either pediatric-inspired or adult-oriented regimens. Recent studies have demonstrated that ‘pediatric’ protocols are more effective for AYA patients [1–3]. However, even with pediatric-inspired regimens, treatment outcomes in the AYA group are worse than in children (overall survival rates of approx. 60% vs. approx. 90% in children) [2]. This implies a need for better understanding of the biology of T-ALL in the AYA group, to develop more efficient treatment strategies. The identification of biological processes specifically deregulated in AYA T-ALL might pave the way toward targeted therapies, e.g., inhibitors of activated signaling pathways [4].

The current knowledge on the molecular characteristics of AYA T-ALL is limited. There are a few studies using next-generation sequencing to characterize the genomic landscape of T-ALL in children and adults [5–7], none of them specifically focusing on young adults. Whole exome sequencing revealed 2.5 times higher mutation burden (somatic protein-altering SNVs and indels) in adult patients (>15 years) than in children and different frequencies of mutations in driver genes between adults and children [5]. RNA-seq revealed a similar number of gene fusions in children (<18 years of age) and adults (>18 years), yet with lower diversity of fusion types among adults than children [6]. These results indicate differences in the biology of T-ALL among age groups.

miRNA expression profiles and the global miRNA transcriptome have already been investigated in T-ALL by several groups, as reviewed by Correia et al. [8], including in our group [9]. However, it has not been investigated thus far in AYA T-ALL. Since miRNAs are implicated in post-transcriptional gene regulation, aberrantly expressed miRNAs may act as oncogenes (overexpressed miRNAs repressing tumor suppressor genes) or as tumor suppressors (underexpressed miRNAs insufficiently repressing oncogenes). miRNAs affecting essential signaling pathways, cell proliferation, and apoptosis contribute to cancer development and progression. Comprehensive investigation of miRNA transcriptome and miRNAs’ involvement in cellular functions might indicate novel mechanisms of leukemia biology, potentially specific to different patient subsets.

We investigated global miRNA expression by small RNA-seq, followed by in silico prediction of miRNA target genes and their involvement in biological pathways and processes. We aimed to identify miRNAs specifically expressed in AYA T-ALL and to focus on those with potential therapeutic implications. Small RNA-seq demonstrated no considerable differences between AYA and pediatric T-ALL. An important finding is the downregulation of hsa-miR-143-3p in pediatric T-ALL patients. We demonstrated its tumor suppressor potential as shown by the restriction of cell proliferation upon overexpression of this miRNA in T-ALL cells in vitro. We also validated direct interaction of hsa-miR-143-3p with KRAS, FGF1 and FGF9, involved in FGFR (fibroblast growth factor receptor) pathway. Although the roles of this miRNA and its target genes in T-ALL biology need more extensive validation, we discussed the implications of our findings in light of the potential application of FGFR inhibition in this leukemia.

2. Results

We investigated miRNA transcriptome in the AYA T-ALL group as compared to pediatric T-ALL and controls (previously published [9]). In total, we analyzed 71 T-ALL cases: 19 AYA (15–39 years of age) and 52 children (<15 years of age), five control samples (normal T-cells of healthy bone marrow donors <18 years of age) and six samples of normal thymocytes (CD34+ and CD4+CD8+). miRNA-seq results of pediatric and AYA patients were analyzed according to the same methodology. Three samples of pediatric T-ALL from our previous study focusing on pediatric T-ALL, were resequenced as ‘inter-experimental calibrators’ in the current study. Additionally, we applied the correction for batch effect, to enable a joint analysis of all miRNA-seq results. Characteristics of samples are presented in Table S1.
2.1. Overview of miRNA Transcriptome in AYA Patients

Small RNA-seq in 19 AYA T-ALL samples resulted in an average number of 10.5 million reads mapped to mature miRNAs/sample (range: 1.7–26.4 million reads) (Table S2). In total, we identified 1244 miRNAs expressed in AYA T-ALL samples (with at least two reads/sample); 1148 known miRNAs and 96 candidate novel miRNAs (not previously reported in miRBase22) (Table S3). The 10 most highly expressed miRNAs in AYA patients included: hsa-let-7a-5p/7c-5p, hsa-let-7f-5p, hsa-miR-92a-3p, hsa-miR-26a-5p, hsa-miR-21-5p, hsa-miR-128-3p, hsa-let-7i-5p, hsa-miR-181a-5p, hsa-let-7g-5p, hsa-miR-16-5p (Figure S1). This list largely resembles that of which we previously described in pediatric T-ALL patients [9].

2.2. miRNA Transcriptome Does Not Clearly Discriminate AYA and Pediatric T-ALL

To identify miRNAs specifically expressed in AYA patients, which could indicate potentially targetable AYA-specific features of leukemia, we investigated miRNAs differentially expressed between AYA vs. pediatric group as well as miRNAs differentially expressed between T-ALL and controls (Figure 1). Hierarchical clustering method showed that discrimination between AYA and pediatric T-ALL, based on the miRNA transcriptome, is weak: samples of AYA and pediatric patients do not group separately under the major branches of the dendrogram. miRNA expression profile differs mostly between T-ALL (both AYA and pediatric samples) and controls (Figure 1). The PCA plot (Figure S2) also shows that miRNA expression profile does not clearly discriminate between AYA and pediatric patients and is largely similar in both age groups as compared to controls in our study.

![Figure 1](image-url). Differentially expressed miRNAs in AYA T-ALL, pediatric T-ALL and control samples. Heatmap and dendrograms of Z-score normalized miRNA expression levels created for miRNAs differentiating T-ALL samples of AYA, pediatric and normal controls (bone marrow T-cells). Rows represent miRNAs; columns represent samples. Dendrograms are based on complete-linkage hierarchical clusterization and Euclidean distances. The fold changes and the abundance of miRNAs (in log scale) are shown on the left side of the plot.
Out of 1244 miRNAs expressed in AYA T-ALL samples, only 88 miRNAs were differentially expressed between AYA and pediatric patients, including 11 candidate novel miRNAs (Table S3). Among differentially expressed miRNAs, three were downregulated in AYA vs. pediatric patients (hsa-miR-4713-5p, hsa-miR-639, and hsa-miR-6724-5p, the latter was also differentially expressed between AYA and controls), and 85 miRNAs were upregulated in AYA vs. pediatric samples. The upregulated miRNAs included six miRNAs, which were not only differentially expressed between AYA vs. pediatric samples but additionally between pediatric vs. control samples (hsa-miR-143-3p, hsa-miR-151a-3p, hsa-miR-4420-3p, hsa-miR-4728-3p, hsa-miR-4749-3p, novel_miRNA_chr17:7306866-7306887).

We further focused on miRNAs that differentiated AYA T-ALL from pediatric T-ALL and simultaneously differentiated T-ALL from the controls. We identified only seven such miRNAs (Table 1), all but one (hsa-miR-6724-5p) showing upregulation in AYA vs. pediatric patients. Of note, only this particular miRNA was simultaneously differing AYA vs. control samples; the remaining six miRNAs were discriminative between AYA vs. pediatric and pediatric vs. controls (Table S3). The low number of miRNAs specifically expressed in the AYA group is in line with the observations from hierarchical clustering, indicating that the miRNA transcriptome is similar in both age groups.

Table 1. miRNAs differentially expressed in AYA T-ALL vs. pediatric T-ALL.

| miRNA ID                   | Average Number of Normalized Reads | LogFC   | p adj |
|----------------------------|-----------------------------------|---------|-------|
| hsa-miR-6724-3p            | 19                                | −4.411  | 0.013 |
| hsa-miR-143-3p             | 6042                              | 1.677   | 0.011 |
| hsa-miR-151a-3p            | 1979                              | 2.482   | 0.002 |
| hsa-miR-4420-3p            | 1.64                              | 4.461   | 0.005 |
| hsa-miR-4728-3p            | 0.26                              | 2.486   | 0.046 |
| hsa-miR-4749-3p            | 0.67                              | 3.787   | 0.005 |
| novel_miRNA_chr17:7306866-7306887 | 0.49                             | 2.696   | 0.029 |

logFC, fold change in log scale (for comparison of AYA vs. pediatric T-ALL); p adj, p value adjusted for multiple testing with Benjamini and Hochberg method with 0.05 significance level.

2.3. Target Prediction and Pathway Analysis Indicate Deregulation of Cancer-Related Signaling Pathways

To investigate the potential involvement of miRNAs differentially expressed between AYA and pediatric T-ALL samples and controls, we performed in silico target prediction for all miRNAs listed in Table 1, with the exclusion of a novel miRNA identified in our miRNA-seq. We used eight miRNA target prediction algorithms (DIANA-microT, ElMMo, MicroCosm, miRanda, miRDB, PicTar, PITA, TargetScan). We also searched three repositories of experimentally validated miRNA:mRNA interactions (miRecords, miRTarBase, DIANA-TarBase). In addition, we searched miR2Disease, PhenomiR, and PharmacomiR, providing data on the potential involvement of miRNAs in diseases and drug response (Table S4).

We identified 3028 genes potentially repressed by miRNAs specifically deregulated in the AYA T-ALL group (Table S4). For 136 target genes, consistently identified by at least three prediction algorithms (Figure 2a), we analyzed their overrepresentation in the processes and pathways, defined by the Gene Ontology, KEGG, and Reactome databases (Figure 2b; Table S4). Genes targeted by miRNAs deregulated in AYA T-ALL were significantly overrepresented in several pathways known to be related to leukemia and cancer in general. These include: Rap1 (Ras-associated protein-1) signaling, PI3K-Akt (phosphatidylinositol 3-kinase) pathway, and FGFR signaling (FGFR1-4), the latter clearly stands out from this analysis. The vast majority of the genes overrepresented in these processes are predicted to be targets of miR-143-3p (Table S4), highlighting the potential importance of this miRNA in the biology of T-ALL cells.
predictions for miRNAs specifically expressed in AYA T-ALL in relation to the number of prediction algorithms used. (b) Processes and pathways potentially affected by miRNAs specifically expressed in AYA T-ALL, revealed by overrepresentation analysis. The plot presents odds ratios for the selected terms, identified using conditional hypergeometric test, with Benjamini and Hochberg correction for multiple testing and 0.05 significance level. The size of the dots (Count) represents the number of genes (predicted targets of the studied miRNAs) involved in a given biological process; the color of the dots represents p value (p adj) adjusted for multiple testing.

Figure 2. miRNA target prediction and pathway analysis. (a) Number of consistent target gene predictions for miRNAs specifically expressed in AYA T-ALL in relation to the number of prediction algorithms used. (b) Processes and pathways potentially affected by miRNAs specifically expressed in AYA T-ALL, revealed by overrepresentation analysis. The plot presents odds ratios for the selected terms, identified using conditional hypergeometric test, with Benjamini and Hochberg correction for multiple testing and 0.05 significance level. The size of the dots (Count) represents the number of genes (predicted targets of the studied miRNAs) involved in a given biological process; the color of the dots represents p value (p adj) adjusted for multiple testing.
2.4. miR-143-3p Affects Proliferation of T-ALL Cells In Vitro

In miRNA-seq, miR-143-3p showed downregulation in T-ALL patients, which we confirmed by RT-qPCR (Appendix A). Prediction algorithms and overrepresentation analysis revealed several oncogenes potentially targeted by this miRNA, including Kras, Fgf1, Fgf9, Itgα6, and Map3k7 (Table S4). Insufficient repression of oncogenes by miR-143-3p might confer growth-promoting effects in T-ALL cells.

To address our hypothesis on the tumor suppressor role of miR-143-3p and oncogenic effects of its downregulation, we selected two T-ALL cell lines (JURKAT and ALL-SIL) showing low endogenous expression of this miRNA and transduced them to overexpress miR-143-3p. In both cell lines, we observed statistically significant reduction of cell growth upon miR-143-3p overexpression as compared to control (Figure 3), indicating the tumor suppressor potential of this miRNA.

![Figure 3](image_url)

**Figure 3.** Effects of overexpression of miR-143-3p in JURKAT and ALL-SIL T-ALL cell lines. Decreased growth of JURKAT (a) and ALL-SIL (b) cells, as demonstrated by CCK8 assay. Cell growth rates were calculated as fold change of OD450 for each examined time point in reference to the starting point (0 h). pCDH EV, empty vector used as a control; *** $p < 0.001$; * $p < 0.05$ (p-value calculated by two-way ANOVA for miR-143-3p expression as independent variable). RT-qPCR results demonstrating overexpression of miR-143-3p in both T-ALL cell lines upon transduction with pCDH miR-143-3p overexpression vector (c).

2.5. miR-143-3p Interacts with FGF1, FGF9 and KRAS

To validate in silico predictions of miR-143-3p target genes, we used luciferase reporter assays. Co-transfection of HEK293T cells with pCDH vector coding miR-143-3p and pmiRGLO plasmids, containing miRNA responsive elements (MREs) in the 3'UTRs of selected target genes (FGF1, FGF9 and KRAS) resulted in the reduction of reporter luciferase activity as compared to controls (co-transfection with pCDH empty vector). These direct interactions of miR-143-3p and its target genes were lost when mutations were introduced into MREs (Figure 4).

![Figure 4](image_url)

**Figure 4.** Cont.
The major finding of the study is the similarity of miRNA transcriptome in pediatric and AYA patients. An additional but potentially important observation is the downregulation of miR-143-3p in T-ALL patients and its putative involvement in leukemia biology as shown by our in silico predictions and basic functional experiments. Among predicted targets of this miRNA, there are several oncogenes significantly overrepresented in signaling pathways of potential importance for T-ALL cells (Figure 2b). miR-143-3p and its target genes (FGF1, FGF9, KRAS) clearly stand out from this overrepresentation analysis (Table S4). Since these genes are implicated in cell proliferation, apoptosis, and stress responses, we hypothesized that insufficient negative regulation over these oncogenes by downregulated miR-143-3p might serve as an oncogenic mechanism in this leukemia. We demonstrated that overexpression of this miRNA in two T-ALL cell lines resulted in the downregulation of miR-143-3p might serve as an oncogenic mechanism in this leukemia. We demonstrated that overexpression of this miRNA in two T-ALL cell lines resulted in the decrease in relative luciferase activity in the presence of miR-143-3p in reference to control (empty vector). For each graph, the predicted interaction sites of miR-143-3p with MREs are shown, with indication of nucleotides mutated in the rescue experiment. (a) Interactions between miR-143-3p and two MREs in KRAS gene. (b) Interaction between miR-143-3p and MREs in FGF1 gene. (c) Interaction between miR-143-3p and MREs in FGF9 gene.

3. Discussion

The tumor suppressor role of miR-143-3p has not been reported thus far in the context of T-ALL. However, it has been suggested in several cancer types, including solid tumors [10], myeloid and B-cell malignancies [11–13]. In acute lymphoblastic leukemia originating from B-cell precursors (BCP-ALL), miR-143 was shown to be downregulated in diagnostic and relapse samples as compared to remission samples [11]. The authors tested the hypothesis on the tumor suppressor role of this miRNA by cell growth assays in REH cells upon miRNA overexpression, yet no impact on growth rate was observed in this particular B-ALL cell line. The downregulation of miR-143 was concluded as a potential biomarker of leukemic cells, but the underlying mechanisms remained unresolved. Similar results were reported in another study in ALL patients, showing downregulation of miR-143 in diagnostic samples and then increasing miRNA levels along with decreasing burden of leukemic cells upon induction treatment [14]. This study focused purely on clinical utility of miRNA expression, and no functional studies were included to investigate the underlying mechanisms. miR143-3p has been shown to be downregulated in colon cancer samples; upregulation of miR143-3p by transfection of HT-29 colon cancer cell line resulted in decreased expression of ERK5 and its upstream activator, KRAS protein [15].
tion of this miRNA in the MCF-7 cell line resulted in decreased expression of ERK5 and MAP3K7 proteins, and decreased cell viability [16].

Our pathway analysis revealed several genes recurrently overrepresented in signaling pathways of potential importance for cell survival, proliferation, and drug resistance, including FGF1, FGF9, KRAS, and MAP3K7, all implicated in FGFR signaling (Figure 2b; Table S4). Three major downstream pathways of FGFR signaling include PI3K-AKT, STAT and Ras-MAPK signaling [17], all three reported to be deregulated in T-ALL. The latter pathway involves KRAS, MAP3K7, and ERK5 (alias MAPK7), all shown to be regulated by miR-143 in various solid tumors, yet not in T-ALL, thus far.

FGFR signaling is crucial in embryonic development and in adult organism, by the involvement in cell proliferation, cell growth, morphogenesis, angiogenesis, tissue repair and metabolism [18]. Dysregulation of the FGFR pathway due to FGFR gene amplification, overexpression, mutations and rearrangements with at least 15 fusion partner genes has been reported in many cancers, including hematological malignancies [19,20]. In BCP-ALL patients, mutations in FGFR1, FGFR2, FGFR3 are rare (<1%), as shown by targeted next-generation sequencing [19]. The frequency of FGFR mutations in T-ALL is also low; FGFR3 and FGFR4 mutated in approx. 5% of patients, as reported by Rokita et al. [21]. In our unpublished whole genome sequencing dataset, mutations of FGFR genes are also infrequent; FGFR2 mutated in 1.5% and FGFR3 in 3% of pediatric T-ALL patients. These data indicate that there are other mechanisms behind upregulation of this pathway. Thus far, there have been no reports on the overactivation of this pathway in T-ALL by aberrantly expressed miRNAs. However, miRNAs have been shown to affect this pathway in other cancer types [22,23].

FGFR pathway is considered as a therapeutic target—several small molecule inhibitors targeting FGFR receptor tyrosine kinases have been already approved or in clinical trials, both in solid tumors and in hematological malignancies (ClinicalTrials.gov identifier: NCT03011372) [24,25]. These include ponatinib (AP24534) a pan-FGFR inhibitor, targeting all members of the FGFR family (FGFR-1, -2, -3, -4). It has been tested in solid tumors and hematological malignancies, including T-ALL, used in combination with PIM inhibitors [20,26–29]. There is also an ongoing trial on the use of another FGFR inhibitor, pemigatinib (INCB054828) in patients with myeloid/lymphoid neoplasms with FGFR1 rearrangement (ClinicalTrials.gov identifier: NCT03011372).

However, oncogenic overactivation of FGFR signaling pathway might not only be due to aberrations affecting the receptor tyrosine kinase (FGFR), but also due to overexpression of the ligands (FGFs, fibroblast growth factors, such as FGF1 or FGF9) or other components of this pathway, downstream of FGFR (such as KRAS).

FGF family proteins exert broad mitogenic and pro-survival activities; in cancers, they have been implicated in tumor growth, invasion and resistance to anticancer therapies [30]. FGFs are abundantly expressed by bone marrow stromal cells and secreted in exosomes, which are then endocytosed by leukemia cells, contributing to resistance to tyrosine kinase inhibitors (TKIs) [31]. Although T-ALL originates in the thymus, FGFR signaling might create a leukemia-protective microenvironment for T-ALL cells infiltrating the bone marrow niche, similarly to the effects of FGFR signaling demonstrated for AML cells [31,32].

KRAS is a member of GTPase superfamily of proteins, acting as an ‘on/off switch’ molecule in the transduction of signals from growth factors; thus, it is involved in several signaling pathways [33]. Oncogenic activation of KRAS is implicated in many different cancers, mainly by mutations in KRAS [34]. However, mutations affecting this proto-oncogene are not the only mechanism of its activation. Recently, KRAS has been shown to be regulated by non-coding RNAs (IncRNAs, miRNAs and circRNAs) in the context of cancer development [35].

Insufficient repression of FGFR pathway components by miR-143-3p might potentially contribute to survival advantage of leukemic cells. Downregulation of this miRNA might serve as a mechanism of oncogenic activation of this pathway, alternative to aberrations
affecting FGFR receptor tyrosine kinases, typically screened to select patients who could benefit from FGFR-inhibiting therapies.

Limitations

Although we showed miR-143-3p as a novel candidate tumor suppressor affecting T-ALL cell proliferation in vitro, the underlying mechanisms need extended validation. It would be important to distinguish if the reduction of proliferation rate of T-ALL cells, observed upon forced overexpression of this miRNA, stems from induction of apoptosis or impairment of proliferation. By dual luciferase reporter assays, we validated in silico predictions of FGF1, FGF9, and KRAS as targets for miR-143-3p. However, these assays demonstrate in vitro interactions of miRNAs with MREs in 3′UTRs of the target genes in an artificial experimental setting. The functional impact of miR-143-3p on deregulation of FGFR signaling in T-ALL cell lines and patients should ideally be verified by a proteomic approach, enabling a global overview of proteins and pathways affected by this miRNA.

4. Materials and Methods

4.1. T-ALL Primary Samples and Controls

T-ALL samples of all patients were obtained at initial diagnosis. Samples of patients aged < 18 years were collected at the centers of Polish Pediatric Leukemia and Lymphoma Study Group. Details of sample preparation, including RNA isolation and quality control were described previously [9]. Samples of patients aged ≥ 18 years were collected in centers of Polish Adult Leukemia Group. Control samples of 5 healthy unrelated bone marrow donors, aged < 18 years were subjected to isolation of mononuclear cells using density gradient centrifugation followed by immunomagnetic separation using Human T Lymphocyte Enrichment Set-DM (Becton Dickinson, Franklin Lakes, NJ, USA) to obtain normal T-cells as controls. Thymocyte CD4+ CD8+ and CD34+ samples, obtained from 3 children undergoing cardiac surgery, were also used as controls. CD34+ thymocyte samples were obtained using MACS purification with CD34 microbeads (Miltenyi Biotec, Bergisch Gladbach, Germany); CD4+ CD8+ thymocytes were obtained by CD4 and CD8 labeling and sorting by a FACSAriaIII (BD Biosciences, Franklin Lakes, NJ, USA), as previously published [36,37]. Characteristics of samples are presented in Table S1. This includes immunophenotypic EGIL classification of T-ALL samples and classification into genetic subtypes based on an oncogenic activation of the following genes TLX1, TLX3, TAL1, TAL2, LMO1, LMO2, LYL1, and HOXA genes. Genetic subtypes were defined based on gene expression levels and fusions identified using RNA-seq (Array Express E-MTAB-11759). RNA-seq results were available for the majority of patients < 18 years at diagnosis and for two patients > 18 years of age (Appendix B).

4.2. T-ALL Cell Culture In Vitro

HEK293T cells were cultured under standard conditions in Dulbecco’s modified Eagle’s medium (Gibco, Thermo Fisher Scientific, Waltham, MA, USA) with 10% fetal bovine serum (Gibco, Thermo Fisher Scientific) and 1% penicillin/streptomycin solution (Sigma Aldrich, St. Louis, MO, USA). T-ALL cells were cultured under standard conditions in RPMI-1640 medium (Gibco, Thermo Fisher Scientific) with 10% or 20% of fetal bovine serum (Gibco, Thermo Fisher Scientific) for JURKAT and ALL-SIL, respectively.

4.3. Small RNA Sequencing and Bioinformatics Analyses

In this study, we used small RNA-seq results of pediatric T-ALL (aged < 18) and control samples, which we obtained previously; data are available in the ArrayExpress database (http://www.ebi.ac.uk/arrayexpress) under accession number E-MTAB-7446 [9]. Patients from this first batch, aged < 15 years, were analyzed as pediatric T-ALL in this study, while patients aged 15–18 years were analyzed in the AYA group. The remaining samples were sequenced in two following batches. The proportion of pediatric and AYA patients in each batch is as follows: batch I (28 pediatric, 6 AYA), batch II (27 pediatric, 7 AYA), batch III
(3 pediatric, 6 AYA). The same 3 pediatric samples were used as 'interexperimental calibra-
tors' in sequencing of all three batches, to enable normalization and correction for batch
effect using edgeR. In the final analyses, these 3 samples (sequenced in 3 batches) were rep-
resented by only one replicate each (samples with the highest normalized read counts were
used). The effectiveness of batch effect correction is presented in PCA plots demonstrating
clustering of samples from all 3 batches before and after correction (Figure S2).

In brief, in batch I, the libraries were generated with NEBNext Multiplex Small RNA
Library Prep Set for Illumina (New England Biolabs, Ipswich, MA, USA), quality controlled
based on size distribution and concentration using 2100 Bioanalyzer (Agilent Technologies,
Santa Clara, CA, USA), and sequenced using NextSeq500 Illumina and standard settings:
10 million reads/per sample, read length: 51 bp single-end. FASTQ files for each sample
were generated using the bcl2fastq software v2.20 (Illumina, San Diego, CA, USA). NGS
service was performed by Exiqon (Exiqon, Vedbæk, Denmark). Sequencing of samples in
batches II and III was conducted at QIAGEN Genomic Services (Düsseldorf, Germany). Li-
braries were prepared using QIAseq miRNA Library Kit, including the ligation of adapters
containing unique molecular identifiers, UMIs (QIAGEN). Quality control of libraries
preparation was performed using 2100 Bioanalyzer (Agilent Technologies, Santa Clara, CA,
USA). Based on the quality of the inserts and the concentration measurements, the libraries
were pooled in equimolar ratios. The library pools were quantified using qPCR. The li-
brary pools were then sequenced on a NextSeq 500 Illumina using standard settings. Raw
data were de-multiplexed, and FASTQ files were generated using the bcl2fastq software
(Illumina, San Diego, CA, USA).

Quality control of reads was conducted using FastQC ver. 0.11.5 (http://www.
biobitinformatics.babraham.ac.uk/projects/fastqc, accessed on 8 January 2019), FastQ Screen
ver. 0.5.1 [38], and custom data visualization scripts. Raw sequencing reads were adapter-
trimmed using Cutadapt [39] (ver. 1.11) and aligned with Bowtie [40] (ver. 1.2.2) to a
modified version of miRBase (ver. 22) created according to the miRge specifications [41].
We used an iterative alignment of reads: reads were first aligned to mature miRNA se-
quences (miRBase ver. 22); unaligned reads were sequentially matched against hairpin
miRNAs (miRBase ver. 22), noncoding RNAs, (Ensembl cDNA database), and again to
mature miRNA sequences (miRBase ver. 22) using less stringent criteria [41]. Detection of
candidate novel miRNAs was based on miRge2 for each individual sample. The results
were later combined between samples based on genomic coordinates of identified miRNAs
(partial overlap was considered sufficient). Both known and candidate novel miRNAs, of
which at least two reads were aligned in a single sample, were further analyzed. Read
normalization and identification of differentially expressed miRNAs, was conducted using
edgeR [42], accounting for batch effect by specifying an appropriate model matrix with
Benjamini and Hochberg correction for multiple testing and 0.05 significance level. Read
counts used in the tests were normalized in edgeR using Trimmed Mean of Mvalues (TMM)
algorithm. If more than two groups existed in a specific classification method, we con-
ducted pairwise comparisons between them for each unique combination and additionally
performed an ANOVA test independently for each miRNA with Benjamini and Hochberg
correction for multiple testing. ANOVA was conducted using edgeR normalized data after
batch effect correction based on ComBat [43]. Results of all analyses were presented as
heatmaps of Z-score normalized miRNA expression levels with dendrograms based on
complete-linkage hierarchical clusterization and Euclidean distances.

4.4. RT-qPCR Validation of miRNAs’ Expression

RT-qPCR validation of expression of hsa-miR-143-3p, hsa-miR-151-3p, hsa-miR-582-5p
and hsa-miR-6724-5p was performed as described previously [9]. Briefly, RNA samples
were reverse transcribed with TaqMan Advanced miRNA cDNA Synthesis Kit (Thermo
Fisher Scientific). TaqMan Fast Advanced Master Mix, predesigned TaqMan Advanced
miRNA assays (Thermo Fisher Scientific), and 7900HT Fast Real-Time PCR System
(Applied Biosystems, Waltham, MA, USA) were used. Relative quantification of expression
was performed using comparative delta CT method ($\Delta \Delta CT$) [44] based on three endogenous normalizer miRNAs (hsa-miR-16-5p, hsa-miR-25-3p and hsa-let-7a-5p) as described previously [45]. Two-tailed Student’s t test was used to test for the significance of differences in expression between T-ALL samples and controls, with $p < 0.05$ for statistical significance.

4.5. miR-143-3p Overexpression in T-ALL Cells In Vitro

For assembly of lentiviral particles, HEK293T cells were seeded on a 6-well culture plate. Upon 70–80% confluence, the cells were transduced with 600 ng of each: PRSV.REV, pMSCV-VSV-G and pMDLg/PRRE packing vectors and 1200 ng of transfer vector. Transfection was performed using JetPrime DNA/siRNA Transfection Kit (Polyplus Transfection, New York, NY, USA). After 24 h, the transfection medium was replaced with 1 mL fresh medium. After 48 h, the medium was collected and filtered with 0.45 µm filters. For transduction, cells were seeded on a 6-well plate in 1.8 mL of RPMI-1640 medium. Then, 200 µL of filtered medium containing lentiviral particles was added to each well. Poli-brene Reagent (Sigma Aldrich, St. Louis, MO, USA) was added to each well at the final concentration of 8 µg/mL. Cells were spinfected for 90 min (at 1500 rpm and 32 °C). Antibiotic selection of transduced cells started 3–4 days post transduction. For selection, puromycin dihydrochloride (Gibco, Thermo Fisher Scientific) was used at the concentration of 10 µg/mL. The selection procedure was conducted for 7 days. The effectiveness of transduction was assessed with the use of CytoFlex S flow cytometer (Beckman Coulter, Indianapolis, IN, USA) with GFP as a marker.

For miR-143-3p overexpression, the pre-miRNA sequence was amplified and cloned into pCDH-CMV-MCS-EF1-GreenPuro overexpression vector (System Biosciences, Frontstraat, The Netherlands). For Dual Luciferase Reporter Assay 3′ UTR sequences coding putative miRNA response element (MRE) for hsa-miR-143-3p in KRAS, FGF1 and FGF9 3′ UTRs, flanked by 30 nt on each side, were cloned into pmiRGLO vector. For rescue experiments, 4 point mutations were introduced to the MRE region during the oligonucleotide synthesis step to abolish the miRNA-mRNA interaction.

4.6. Proliferation Assays

After transduction and puromycin selection, $2 \times 10^4$ cells per well were seeded on 96-well plate in 100 µL of culture medium. After 0, 24, 48, 72, and 96 h, 10 µL of Cell Counting Kit 8 (Sigma Aldrich, Bornem, Belgium) reagent was added to each well. Cells were incubated with CCK8 reagent for 4 h in 37 °C. Absorbance was measured on the GloMax-Multi+ Detection System (Promega, Madison, WI, USA). Each experiment was conducted in three technical and three biological replicates. The growth rate for each replicate in each time point was calculated as a fold change of absorbance in the first tested time point (0 h). Statistical significance was calculated with two-way ANOVA test or miR-143-3p expression as independent variable.

4.7. Dual Luciferase Reporter Assays

Predicted miRNA-mRNA interactions were validated with Dual-Glo Luciferase Reporter Assay (Promega, Madison, WI, USA). HEK 293T cells were seeded on 24-well culture plate 24 h before transfection. Cells were subjected to transfection at 60% to 80% confluency using JetPrime DNA/siRNA Transfection Kit (Polyplus Transfection, New York, NY, USA) to enable co-transfection with miR-143-3p coding or empty pCDH vector and pmiRGLO plasmids (Promega), containing 3′UTRs of the selected target genes. 250 ng and 50 ng of pCDH and pmiRGLO plasmid, respectively, was added per well. Luciferase activity was measured with GloMax-Multi+ Detection System (Promega) after 72 h from transfection. All experiments were performed in four replicates. A significant decrease in luciferase activity relative to control was indicative of direct interaction between the seed sequence of the miRNA (defined as the nucleotides at position 2–7 of the 5′ end of mature miRNA sequences) and the MRE in the 3′ UTR of target mRNA.
5. Conclusions

miRNA-seq failed to reveal major differences between AYA and pediatric T-ALL patients. This is in line with the reports from other high-throughput analyses, indicating that the landscape of aberrations in AYA cases is at the interphase of those observed in pediatric and adult patients. The integration of multiomics data obtained in AYA patients seems to be another step to take. This approach might prove more powerful for the characterization of the molecular landscape of this leukemia in AYA patients and, more importantly, for the identification of AYA-specific, therapeutically targetable features. However, the heterogeneity of many cancers, including T-ALL, which is increasingly evident with the cumulation of omics and multiomics data, indicates another possible direction in the development of personalized therapy. The aim should be the identification of subgroups of patients, sharing the same molecular features (e.g., dysregulated signaling pathway), who might benefit from the same targeted therapy, regardless of the patients age, immunophenotypic or histological subtype and even regardless of the type of cancer diagnosed. The identification of patient subsets sharing the same driver aberrations and targetable lesions will enable drug repurposing and development of combination therapeutic strategies. In this respect, the identification of miRNAs contributing to the dysregulation of key signaling pathways is a line of basic research with potential future implications for the development of novel therapeutic options.

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Informed Consent Statement: Informed consent of patients/legal guardians was obtained from subjects involved in the study.

Data Availability Statement: Small RNA-seq results of samples from batch I (T-ALL patients aged < 18 and controls) previously published [9] is available in the ArrayExpress database (http://www.ebi.ac.uk/arrayexpress) under accession number E-MTAB-7446. Small RNA-seq results of samples from batches II and III (pediatric and AYA patients) are available in the ArrayExpress database under accession number E-MTAB-11987.

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Appendix A. RT-qPCR Validation of miRNA-Seq Results

For RT-qPCR validation, we selected three miRNAs based on the highest average read counts in miRNA-seq, namely: hsa-miR-143-3p, hsa-miR-151-3p, and hsa-miR-6724-5p. The expression levels of the remaining four miRNAs were low, precluding their detection by RT-qPCR. For hsa-miR-151-3p, RT-qPCR failed to validate our miRNA-seq results. For hsa-miR-6724-5p, we showed its downregulation in AYA vs. controls yet failed to confirm its lower expression in AYA vs. pediatric patients. In the case of miR-143-3p, we confirmed that it is highly expressed in AYA vs. pediatric samples, and we showed that its expression is downregulated in both age groups of T-ALL patients as compared to controls, indicating its potential tumor suppressor role in T-ALL.

Figure A1. RT-qPCR validation of miRNA-seq results: * p < 0.05; ** p < 0.01; *** p < 0.001; ns (not significant).

Appendix B. miRNAs Differentially Expressed in T-ALL Subtypes

Patients with available data on the immunophenotype of leukemic cells at diagnosis were classified into four immunophenotypic EGIL subtypes. miRNAs differentially expressed between EGIL subtypes are presented in Table S3. Patients with available data from mRNA-seq were classified into genetic subtypes based on aberrant expression/fusion of driver oncogenes. Patients characteristics is presented in Table S1. The PCA plot based on mRNA-sequencing showed grouping of patients into three clusters C1-C3.

Figure A2. PCA plot based on mRNA sequencing presenting grouping of samples into clusters.
Cluster C3 is mainly composed of samples classified into the TAL1 subtype, and it includes all three samples showing NKX2 expression. Cluster C2 is mainly composed of samples from the TLX3 subtype, while cluster C1 is enriched in LMO1, LMO2, LYL1 and HOXA cases. miRNAs differentially expressed between clusters C1, C2, C3, enriched in specific genetic subtypes are presented in Table S3.

References

1. Muffly, L.; Alvarez, E.; Lichtensztajn, D.; Abrahão, R.; Gomez, S.L.; Keegan, T. Patterns of care and outcomes in adolescent and young adult acute lymphoblastic leukemia: A population-based study. *Blood Adv.* 2018, 2, 895–903. [CrossRef] [PubMed]

2. Testi, A.M.; Canichella, M.; Vitale, A.; Piciocchi, A.; Guarini, A.; Starza, I.D.; Cavalli, M.; De Propris, M.S.; Messina, M.; Elia, L.; et al. Adolescent and young adult acute lymphoblastic leukemia. Final results of the phase II pediatric-like GIMEMA LAL-1308 trial. *Am. J. Hematol.* 2021, 96, 292–301. [CrossRef] [PubMed]

3. White, V.M.; Skaczkowski, G.; Pinkerton, R.; Coory, M.; Osborn, M.; Bibby, H.; Nicholls, W.; Orme, L.M.; Conyers, R.; Phillips, M.B.; et al. Clinical management of Australian adolescents and young adults with acute lymphoblastic and myeloid leukemias: A national population-based study. *Pediatr. Blood Cancer* 2018, 65, e27349. [CrossRef] [PubMed]

4. Friend, B.D.; Schiller, G.J. Closing the gap: Novel therapies in treating acute lymphoblastic leukemia in adolescents and young adults. *Blood Rev.* 2018, 32, 122–129. [CrossRef]

5. De Keersmaecker, K.; Atak, Z.K.; Li, N.; Vicente, C.; Patchett, S.; Girardi, T.; Gianfelici, V.; Geerdens, E.; Clappier, E.; Porcu, M.; et al. Exome sequencing identifies mutation in CNOT3 and ribosomal genes RPL5 and RPL10 in T-cell acute lymphoblastic leukemia. *Nat. Genet.* 2013, 45, 186–190. [CrossRef]

6. Chen, B.; Jiang, L.; Zhong, M.L.; Li, J.F.; Li, B.S.; Peng, L.J.; Dai, Y.T.; Cui, B.W.; Yan, T.Q.; Zhang, W.N.; et al. Identification of fusion genes and characterization of transcriptome features in T-cell acute lymphoblastic leukemia. *Proc. Natl. Acad. Sci. USA* 2018, 115, 373–378. [CrossRef]

7. Liu, Y.; Easton, J.; Shao, Y.; Maciaszek, J.; Wang, Z.; Wilkinson, M.R.; McCastlain, K.; Edmonson, M.; Pounds, S.B.; Shi, L.; et al. The genomic landscape of pediatric and young adult T-lineage acute lymphoblastic leukemia. *Nat. Genet.* 2017, 49, 1211–1218. [CrossRef]

8. Correia, N.C.; Barata, J.T. MicroRNAs and their involvement in T-ALL: A brief overview. *Adv. Biol. Regul.* 2019, 74, 100650. [CrossRef]
10. Dawidowska, M.; Jaksik, R.; Drobn, M.; Szarzyńska-Zawadzka, B.; Kosmalska, M.; Sędę, L.; Machowska, L.; Lalik, A.; Lejman, M.; Ussowiec, M.; et al. Comprehensive Investigation of miRNome Identifies Novel Candidate miRNA-mRNA Interactions Implicated in T-Cell Acute Lymphoblastic Leukemia. *Neoplasia* 2019, 21, 294–310. [CrossRef]

11. Noguchi, S.; Mori, T.; Hoshino, Y.; Maruo, K.; Yamada, N.; Kitade, Y.; Naoe, T.; Aka, Y. MicroRNA-143 functions as a tumor suppressor in human bladder cancer T24 cells. *Cancer Lett.* 2011, 307, 211–220. [CrossRef]

12. Yang, Y.-L.; Jou, S.-T.; Lin, S.-W.; Lin, D.-T.; Hu, C.-Y., Jr.; Chang, S.-K.; Yen, C.-T.; Chiou, R.-J.; Lin, K.-H.; Lu, M.-Y.; et al. Down-Regulation of MicroRNA-143 and -145 in Childhood B-Lineage Acute Lymphoblastic Leukemia at Initial Diagnosis and in Relapse but up-Regulated When in Remission. *Blood* 2008, 112, 4886. [CrossRef]

13. Hartmann, J.U.; Bräuer-Hartmann, D.; Kardosova, M.; Wurm, A.A.; Willke, F.; Schödel, C.; Gerloff, D.; Katzerke, C.; Krakowsky, R.; Namasu, C.Y.; et al. MicroRNA-143 targets ERK5 in granulopoiesis and predicts outcome of patients with acute myeloid leukemia. *Cell Death Dis.* 2018, 9, 814. [CrossRef] [PubMed]

14. Shen, J.Z.; Zhang, Y.Y.; Fu, H.Y.; Wu, D.S.; Zhou, H.R. Overexpression of microRNA-143 inhibits growth and induces apoptosis in human leukemia cells. *Oncol. Rep.* 2014, 31, 2035–2042. [CrossRef] [PubMed]

15. Piatopoulou, D.; Avgeris, M.; Drakaki, I.; Marmarinos, A.; Xagorari, M.; Baka, M.; Poursidis, A.; Kossiva, L.; Gourgiotis, D.; Scorilas, A. Clinical utility of miR-143/miR-182 levels in prognosis and risk stratification specificity of BFM-treated childhood acute lymphoblastic leukemia. *Ann. Hematol.* 2015, 94, 1169–1182. [CrossRef] [PubMed]

16. Pekow, J.; Meckel, K.; Dougherty, U.; Butun, F.; Mustafi, R.; Lim, J.; Crofton, C.; Chen, X.; Joseph, L.; Bissonnette, M. Tumor suppressors miR-143 and miR-145 and predicted target proteins API5, ERK5, K-RAS, and IRS-1 are differentially expressed in proximal and distal colon. *Am. J. Pathol.* 2015, 308, G179–G187. [CrossRef]

17. Zhou, L.L.; Dong, J.L.; Huang, G.; Sun, Z.L.; Wu, J. MicroRNA-143 inhibits cell growth by targeting ERK5 and MAPK7 in breast cancer. *Braz. J. Med. Biol. Res.* 2017, 50, e5891. [CrossRef]

18. Szybowska, P.; Kostas, M.; Węsche, J.; Haugsten, E.M.; Wiedlocha, A. Negative Regulation of FGFR (Fibroblast Growth Factor Receptor) Signaling. *Cells* 2021, 10, 1342. [CrossRef] [PubMed]

19. Xiao, J.F.; Caliri, A.W.; Duex, J.E.; Theodosiou, D. Targetable Pathways in Advanced Bladder Cancer: FGFR Signaling. *Cancers* 2021, 13, 4891. [CrossRef]

20. Jerchel, I.S.; Hoogkamer, A.Q.; Ariës, I.M.; Boer, J.M.; Besselink, N.J.M.; Koudijs, M.J.; Pieters, R.; den Boer, M.L. Fibroblast growth factor receptor signaling in pediatric B-cell precursor acute lymphoblastic leukemia. *Sci. Rep.* 2019, 9, 1875. [CrossRef]

21. Carll, T.; Patel, A.; Derman, B.; Hyjek, E.; Lager, A.; Wanjari, P.; Segal, J.; Odenike, O.; Fidai, S.; Arber, D. Diagnosis and treatment of mixed phenotype (T/myeloid/lymphoid) acute leukemia with novel ETV6-FGFR2 rearrangement. *Blood Adv.* 2020, 4, 4924–4928. [CrossRef]

22. Rokita, J.L.; Rathii, K.S.; Cardenas, M.F.; Upton, K.A.; Jayaseelan, J.; Cross, K.L.; Pfeil, J.; Egolf, L.E.; Way, G.P.; Farrel, A.; et al. Genomic Profiling of Childhood Tumor Patient-Derived Xenograft Models to Enable Rational Clinical Trial Design. *Cell Rep.* 2019, 29, 1675–1689.e1679. [CrossRef] [PubMed]

23. Zhang, J.; Zhou, Y.; Huang, T.; Wu, F.; Pan, Y.; Dong, Y.; Wang, Y.; Chan, A.K.Y.; Liu, L.; Kwan, J.S.H.; et al. FGFR18, a prominent player in FGFR signaling, promotes gastric tumorigenesis through autocrine manner and is negatively regulated by miR-590-5p. *Oncogene* 2019, 38, 33–46. [CrossRef]

24. Yang, Y.; Zhang, J.; Xia, T.; Li, G.; Tian, T.; Wang, M.; Wang, R.; Zhao, L.; Lan, K.; Zhou, W. MicroRNA-210 promotes cancer angiogenesis by targeting fibroblast growth factor receptor-1 like 1 in hepatocellular carcinoma. *Oncol. Rep.* 2016, 36, 2553–2562. [CrossRef] [PubMed]

25. Dai, S.; Zhou, Z.; Chen, Z.; Xu, G.; Chen, Y. Fibroblast Growth Factor Receptors (FGFRs): Structures and Small Molecule Inhibitors. *Cells* 2019, 8, 614. [CrossRef]

26. Chae, Y.K.; Ranganath, K.; Hammond, P.S.; Vaklavas, C.; Mohindra, N.; Kalyan, A.; Matsangou, M.; Costa, R.; Carneiro, B.; Villaflor, V.M.; et al. Inhibition of the fibroblast growth factor receptor (FGFR) pathway: The current landscape and barriers to clinical application. *Oncotarget* 2017, 8, 16052–16074. [CrossRef]

27. Greulich, H.; Pollock, P.M. Targeting mutant fibroblast growth factor receptors in cancer. *Trends Mol. Med.* 2011, 17, 283–292. [CrossRef] [PubMed]

28. Musumeci, F.; Greco, C.; Grossi, G.; Molinari, A.; Schenone, S. Recent Studies on Ponatinib in Cancers Other Than Chronic Myeloid Leukemia. *Cancers* 2018, 10, 430. [CrossRef] [PubMed]

29. Padi, S.K.R.; Luevano, L.A.; An, N.; Pandey, R.; Singh, N.; Song, J.H.; Aster, J.C.; Yu, X.Z.; Mehrotra, S.; Kraft, A.S. Targeting the PIM protein kinases for the treatment of a T-cell acute lymphoblastic leukemia subset. *Oncotarget* 2017, 8, 30199–30216. [CrossRef]

30. Khodadoust, M.S.; Luo, B.; Medeiros, B.C.; Johnson, R.C.; Ewalt, M.D.; Schalkwyk, A.S.; Bangs, C.D.; Cherry, A.M.; Arai, S.; Arber, D.A.; et al. Clinical activity of ponatinib in a patient with FGFR1-rearranged mixed-phenotype acute leukemia. *Leukemia* 2016, 30, 947–950. [CrossRef] [PubMed]

31. Yun, Y.R.; Won, J.E.; Jeon, E.; Lee, S.; Kang, W.; Jo, H.; Jang, J.H.; Shin, U.S.; Kim, H.W. Fibroblast growth factors: Biology, function, and application for tissue regeneration. *J. Tissue Eng.* 2010, 2010, 218142. [CrossRef]

32. Javidi-Sharifi, N.; Martinez, J.; English, I.; Joshi, S.K.; Scopim-Ribeiro, R.; Viola, S.K.; Edwards, D.K.; Agarwal, A.; Lopez, C.; Jorgens, D.; et al. FGFR2-FGFR1 signaling regulates release of Leukemia-Protective exosomes from bone marrow stromal cells. *Elife* 2019, 8, e40033. [CrossRef] [PubMed]
32. Calvo, J.; Fahy, L.; Uzan, B.; Pflumio, F. Desperately seeking a home marrow niche for T-cell acute lymphoblastic leukaemia. *Adv. Biol. Regul.* 2019, 74, 100640. [CrossRef]

33. Zhu, Z.; Golay, H.G.; Barbie, D.A. Targeting pathways downstream of KRAS in lung adenocarcinoma. *Pharmacogenomics* 2014, 15, 1507–1518. [CrossRef]

34. Huang, L.; Guo, Z.; Wang, F.; Fu, L. KRAS mutation: From undruggable to druggable in cancer. *Signal Transduct. Target* 2021, 6, 386. [CrossRef] [PubMed]

35. Ghafouri-Fard, S.; Shirvani-Farsani, Z.; Hussien, B.M.; Taheri, M.; Jalili Khoshnoud, R. Emerging role of non-coding RNAs in the regulation of KRAS. *Cancer Cell Int.* 2022, 22, 68. [CrossRef]

36. Taghon, T.; Waegemans, E.; Van de Walle, I. Notch signaling during human T cell development. *Curr. Top. Microbiol. Immunol.* 2012, 360, 75–97. [CrossRef] [PubMed]

37. Wallaert, A.; Van Loocke, W.; Hernandez, L.; Taghon, T.; Speleman, F.; Van Vlierbergh, P. Comprehensive miRNA expression profiling in human T-cell acute lymphoblastic leukemia by small RNA-sequencing. *Sci. Rep.* 2017, 11, 7–7901. [CrossRef] [PubMed]

38. Wingett, S.W.; Andrews, S. FastQ Screen: A tool for multi-genome mapping and quality control. *F1000 Res.* 2018, 7, 1338. [CrossRef]

39. Marcel, M. Cutadapt Removes Adapter Sequences From High-Throughput Sequencing Reads. *EMBnet J.* 2012, 17, 10–12. [CrossRef]

40. Langmead, B.; Trapnell, C.; Pop, M.; Salzberg, S.L. Ultrafast and memory-efficient alignment of short DNA sequences to the human genome. *Genome Biol.* 2009, 10, R25. [CrossRef]

41. Baras, A.S.; Mitchell, C.J.; Myers, J.R.; Gupta, S.; Weng, L.C.; Ashton, J.M.; Cornish, T.C.; Pandey, A.; Halushka, M.K. miRge—A Multiplexed Method of Processing Small RNA-Seq Data to Determine MicroRNA Entropy. *PLoS ONE* 2015, 10, e0143066. [CrossRef]

42. Robinson, M.D.; McCarthy, D.J.; Smyth, G.K. edgeR: A Bioconductor package for differential expression analysis of digital gene expression data. *Bioinformatics* 2010, 26, 139–140. [CrossRef] [PubMed]

43. Johnson, W.E.; Li, C.; Rabinovic, A. Adjusting batch effects in microarray expression data using empirical Bayes methods. *Biostatistics* 2007, 8, 118–127. [CrossRef] [PubMed]

44. Schmittgen, T.D.; Livak, K.J. Analyzing real-time PCR data by the comparative C(T) method. *Nat. Protoc.* 2008, 3, 1101–1108. [CrossRef]

45. Drobna, M.; Szarzyńska-Zawadzka, B.; Daca-Roszak, P.; Kosmalska, M.; Jaksik, R.; Witt, M.; Dawidowska, M. Identification of Endogenous Control miRNAs for RT-qPCR in T-Cell Acute Lymphoblastic Leukemia. *Int. J. Mol. Sci.* 2018, 19, 2858. [CrossRef] [PubMed]