



# Advantages and Limitations of SNP Array in the Molecular Characterization of Pediatric T-Cell Acute Lymphoblastic Leukemia

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### Specialty section:

This article was submitted to  
Pediatric Oncology,  
a section of the journal  
Frontiers in Oncology

**Received:** 17 March 2020

**Accepted:** 11 June 2020

**Published:** 17 July 2020

### Citation:

Lejman M, Włodarczyk M, Styka B, Pastorczak A, Zawitkowska J, Taha J, Sędek Ł, Skonieczka K, Braun M, Haus O, Szczepański T, Młynarski W and Kowalczyk JR (2020) Advantages and Limitations of SNP Array in the Molecular Characterization of Pediatric T-Cell Acute Lymphoblastic Leukemia. *Front. Oncol.* 10:1184. doi: 10.3389/fonc.2020.01184

T-cell acute lymphoblastic leukemia (T-ALL) is a highly heterogeneous disease, and numerous genetic aberrations in the leukemic genome are responsible for the biological and clinical differences among particular ALL subtypes. However, there is limited knowledge regarding the association of whole-genome copy number abnormalities (CNAs) in childhood T-ALL with the course of leukemia and its outcome. The aim of this study was to identify the pattern of whole-genome CNAs in 86 newly diagnosed childhood T-ALL cases using a high-density single-nucleotide polymorphism array. We analyzed the presence of whole-genome CNAs with respect to immunophenotype, clinical features, and treatment outcomes. A total of 769 CNAs, including trisomies, duplications, deletions, and segmental loss of heterozygosity, were detected in 86 analyzed samples. Gain or loss of chromosomal regions exceeding 10Mb occurred in 46 cases (53%), including six cases (7%) with complex chromosomal alterations. We observed that microdeletions in selected genes (e.g., *FIP1L1* and *PDGFRB*) were related to the clinical features. Interestingly, 13% of samples have a duplication of the two loci (*MYB* and *AIH1*—6q23.3), which never occurred alone. Single-nucleotide polymorphism array significantly improved the molecular characterization of pediatric T-ALL. Further studies with larger cohorts of patients may contribute to the selection of prognostic CNAs in this group of patients.

**Keywords:** childhood, T-cell acute lymphoblastic leukemia, SNP array, CNAs, molecular characterization

## INTRODUCTION

T-cell acute lymphoblastic leukemia (T-ALL) accounts for ~15% of pediatric ALL cases (1). T-cell acute lymphoblastic leukemia is a biologically heterogeneous malignancy with numerous genetic aberrations in the leukemic genome. T-cell acute lymphoblastic leukemia has been divided into four subtypes according to the European Group for the Immunological Classification of Leukemia

(EGIL): pro-T EGIL T-I (cCD3<sup>+</sup>, CD7<sup>+</sup>), pre-T EGIL T-II (cCD3<sup>+</sup>, CD7<sup>+</sup>, and CD5/CD2<sup>+</sup>), cortical T EGIL T-III (cCD3<sup>+</sup>, Cd1a<sup>+</sup>, sCD3<sup>+/-</sup>), and mature-T EGIL T-IV (cCD3<sup>+</sup>, sCD3<sup>+</sup>, CD1a<sup>-</sup>) (2). Early T-cell precursor (ETP)-ALL is characterized as an additional subtype of T-cell ALL with blasts usually negative for CD1a and CD8, weak expression of CD5, and the presence of one or more myeloid or stem cell markers (3, 4). The prognostic factors and risk stratification are different in T-ALL when compared to B-lineage leukemia. Minimal residual disease (MRD) response plays an essential role in the risk group assignment; leukocytosis and age at diagnosis are not independent prognostic factors in T-ALL (5). Generally, pre-T ALL shows a worse outcome than pre-B ALL in both pediatric and adult patients (6). Currently, risk-adapted therapy and appropriate supportive care result in a relatively high 5 year overall survival rate of 80%. However, effective treatment of T-ALL relapses is still clinically challenging (5). Because most patients with relapse were originally stratified into an intermediate risk group, none of the existing prognostic genetic markers were efficient enough to predict treatment outcome in T-ALL. In the last 10 years, advanced genomic and transcriptomic studies using high-throughput techniques have provided a better understanding of the pathogenesis of T-ALL (7, 8). Moreover, associations of molecular lesions with specific T-ALL subtypes, clinical features, and outcomes have been documented (9).

However, there is still limited knowledge about the association of whole-genome copy number abnormalities (CNAs) in childhood T-ALL with disease course and outcome. In this study, we analyzed the pattern of whole-genome CNAs using a high-density single-nucleotide polymorphism (SNP) array in childhood ALL, depending on the maturation state of leukemic cells, clinical features, and treatment outcome.

## MATERIALS AND METHODS

### Patients

A series of bone marrow aspirates from 86 children with newly diagnosed T-ALL were analyzed prior to any oncological treatment. All patients were treated in 13 centers of the Polish Pediatric Leukemia/Lymphoma Study Group, according to the following protocols: ALL IC BFM 2002 [ $n = 15$  (17%)], ALL IC BFM 2009 [ $n = 56$  (66%)], and AIEOP BFM ALL 2017 [ $n = 15$  (17%)], within the randomized trial of the International Berlin-Frankfurt-Munster Study Group (I-BFM-SG) for the therapy of childhood ALL. Bone marrow samples from all patients were cytogenetically investigated by G-banding analysis at the time of diagnosis. Targeted fluorescence *in situ* hybridization (FISH) *t*(9;22) (q34; q11)/*BCR-ABL1*, 11q23/*KMT2A* rearrangements were performed in 86 cases (100%). Genetic tests were performed for all ALL samples. The majority of patient samples (49 of 86) were examined in the Department of Diagnostic Genetics of Medical University of Lublin, and these samples were collected from the Department of Hematology, Oncology, and Transplantology in Lublin. The remaining data on patient cytogenetics were previously established within the therapeutic program through a combined cytogenetics FISH screening in the National Reference Center: Department of Clinical Genetics,

Collegium Medicum in Bydgoszcz. Immunophenotypic data enabled the classification of patients into pre-T ALL [ $n = 24$  (28%)], cortical T-ALL [ $n = 23$  (26.6%)], mature T-ALL [ $n = 21$  (24%)], and ETP-ALL [ $n = 2$  (2%)] based on the EGIL criteria. In 16 cases (19%), immunophenotypic data were not available to determine a specific subtype of T-ALL. The clinical and laboratory details of patients with T-ALL are presented in **Table 1**. All of the participants have written consent to participate and publish the data. All procedures performed in studies involving human participants were in accordance with the ethical standards of the institutional and/or national research committee and with the 1964 Helsinki Declaration and its later amendments or comparable ethical standards. Informed consent was obtained from all individuals included in the study and their parents or guardians on behalf of any participant younger than 16 years. The study was approved by the ethics committee (KNW/0022/KB1/153/I/16/17).

### Laboratory Methods

Bone marrow samples were aspirated into anticoagulant (EDTA)-containing tubes from patients with T-ALL at the time of diagnosis. Genomic DNA was isolated with the QIAamp DNA Blood Mini Kit (Qiagen, Hilden, Germany). DNA samples were stored at  $-20^{\circ}\text{C}$  until SNP arrays experiments were performed. The median leukemic cell counts in the collected bone marrow were assessed using flow cytometry to be 82.43% (range = 30–100%). The concentration and quality of isolates were determined by spectrophotometry (NanoDrop 8000; Thermo Fisher Scientific, Waltham, MA, USA). The microarray analyses were performed with the use of a CytoScan HD array [2,670,000 markers, including 750,000 SNP and 1,900,000 non-polymorphic copy number variant (CNV) markers] (Applied Biosystems, part of Thermo Fisher Scientific). All laboratory procedures were carried out according to the manufacturer's protocols. A previously described microarray method that has been standardized in our laboratory based on AIEOP-BFM array screening strategy recommendations was used for microdeletion and microduplication assessments (10). In total, 250 ng of genomic DNA was analyzed in accordance with the manufacturer's protocols. The study was based on an analysis of scanned data files that were generated with the Chromosome Analysis Suite v 3.3 (ChAS; Thermo Fisher Scientific). Furthermore, the copy numbers of altered regions (CNAs) were calculated, and the data were normalized to a reference model (Thermo Fisher Scientific) of baseline reference intensities, NA 33 (hg19/CRCh37). The copy number states and their breakpoints were determined with the use of the hidden Markov model software package. The threshold levels of  $\log_2$  ratio  $\geq 0.5$  and  $\leq 0.5$  were used for the categorization of the altered chromosomal regions as CNV gains and losses, respectively. The identification of normal diploid markers in the cancer samples constituted an essential part of the algorithm, which was particularly significant in highly sample-induced aberrations. Furthermore, unaltered diploid markers were used for the calibration of signals, resulting in a  $\log_2$  ratio of 0 (e.g., copy number 2). The algorithm also indicated that the identified unaltered diploid markers corresponded with

**TABLE 1** | Clinical and biological characteristics of the study group with respect to the implemented treatment protocol.

		ALL IC-BFM 2002 <i>n</i> = 15		ALL IC-BFM 2009 <i>n</i> = 56		AIEOP BFM ALL 2017 <i>n</i> = 15		<i>p</i> -value
Age at diagnosis (years)		9.34 (2.84–17.53)		7.68 (1.08–16.34)		10.06 (2.41–17.67)		0.047
Platelets		100,857.1 (12–000)		110,743.9 (273–490,000)		603,42.21 (60–520,000)		0.057
Hb [g/dl]		9.72 (4.30–15.8)		9.94 (4.7–15.0)		10.84 (7.7–13.0)		0.151
Blasts—peripheral blood (%)		16.16 (0.00–96.0)		17.49 (0.0–96)		20.72 (3.0–80.0)		0.595
Blasts—bone marrow (%)		79.57 (36.0–100)		82.35 (80.0–100)		79.84 (30.0–100)		0.336
WBC count (n/μL)		127,863 (220–791,490)		142,861 (8,687–791,490)		100,144 (220–403,160)		0.490
Sex	Female	4	27%	17	30%	6	40%	0.705
	Male	11	73%	39	70%	9	60%	
Poor steroid response	No	5	33%	32	57%	12	80%	0.155
	Yes	10	67%	24	43%	3	20%	
MRD FMC 15 (%)	<10	10	67%	32	58%	9	60%	0.838
	>10	5	33%	23	42%	6	40%	
Hyperdiploidy	No	13	87%	55	98%	14	93%	0.156
	Yes	2	13%	1	2%	1	7%	
<i>KMT2A</i> +	No	13	87%	52	93%	14	93%	0.710
	Yes	2	13%	4	7%	1	7%	
Karyotype	Normal	10	67%	31	62%	12	80%	0.124
	Complex	4	27%	19	38%	3	20%	
	Hypertriploidy	1	7%	0	0%	0	0%	
Mediastinal tumor	No	8	57%	33	59%	6	43%	0.553
	Yes	6	43%	23	41%	8	57%	
Hepatomegaly	No	4	27%	22	39%	5	33%	0.755
	Yes	11	73%	34	61%	10	67%	
Splenomegaly	No	3	20%	22	39%	4	27%	0.398
	Yes	12	80%	34	61%	11	73%	
Lymph node involvement	No	11	79%	44	79%	13	93%	0.462
	Yes	3	21%	12	21%	1	7%	
Risk group	Non-HR	9	64%	26	47%	8	53%	0.761
	HR	5	36%	29	53%	7	47%	
HSCT	No	10	71%	35	70%	14	93%	0.181
	Yes	4	41%	15	30%	1	7%	
Relapse	No	12	86%	44	86%	14	93%	0.749
	Yes	2	14%	7	14%	1	7%	
Follow up (years)		9.32		3.87		0.63		<b>0.001</b>

WBC, white blood count; MRD FMC, minimal residual disease flow cytometry; HSCT, hematopoietic stem cell transplantation; HR, high-risk group; non-HR, non-high-risk group. Data are presented as medians and IQRs or numbers and percentages. The follow-up time did not differ significantly between the two protocols, ALL IC BFM 2002 (9.32 years) and ALL IC BFM 2009 (8.37 years), but was shorter for AIEOP BFM ALL 2017 ( $p < 0.01$ ). Apart from age at diagnosis, no other significant differences with respect to the clinical and biological features between protocols were observed. Differences between protocols were assessed using the Kruskal–Wallis or the  $\chi^2$  tests. The bold value is considered statistically significant.

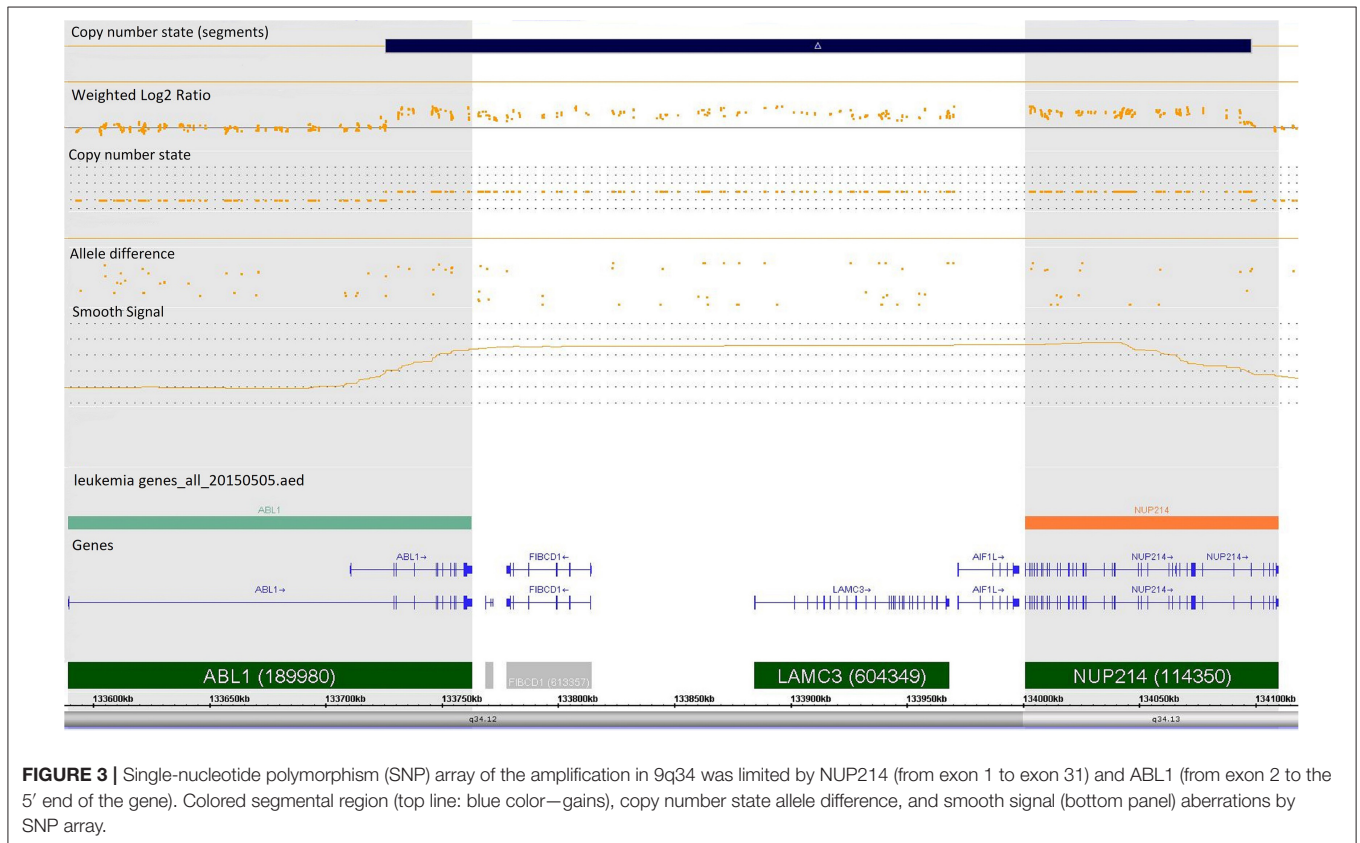
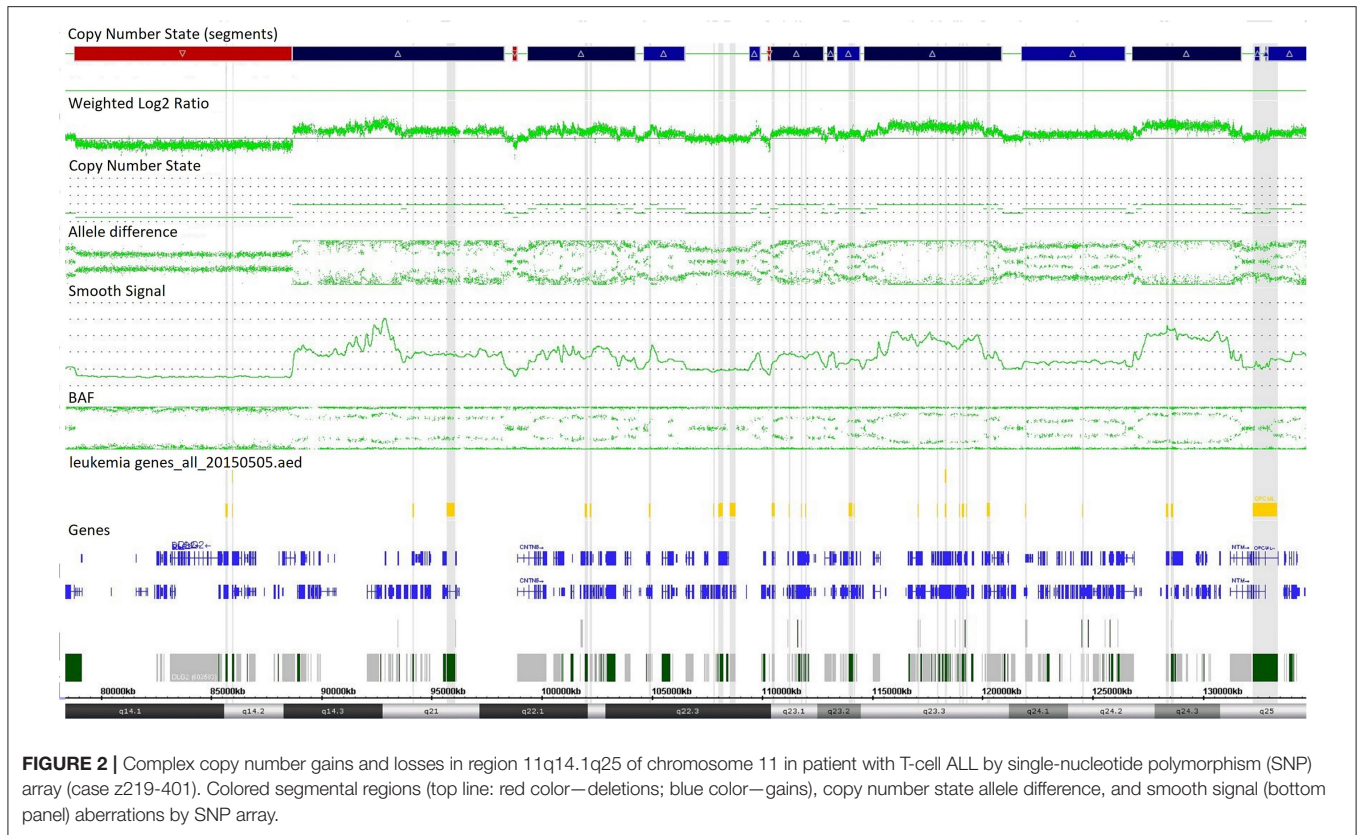
CN = 4. In such a case, the  $\log_2$  ratio was readjusted, and the chromosomal ploidy of 4 was reported. The obtained data were analyzed based on two different criteria: genome-wide CNVs and leukemia-associated region/gene-specific CNAs (leukemia genes\_all\_20150505; Fullerton Overlap Map\_hg19). Our panel of 92 selected genes was created based on comprehensive literature describing genetic alterations and their role in pathogenesis T-ALL (7, 9, 11, 12). The genes are listed in **Table S1**. The minimal number of probes was applied to determine the CNAs were 50 probes for duplication (gain) and 25 probes for deletion (loss). Copy number–neutral loss of heterozygosity (LOH) is reported when the length is >3,000 kbp. To further identify the genes involved in the

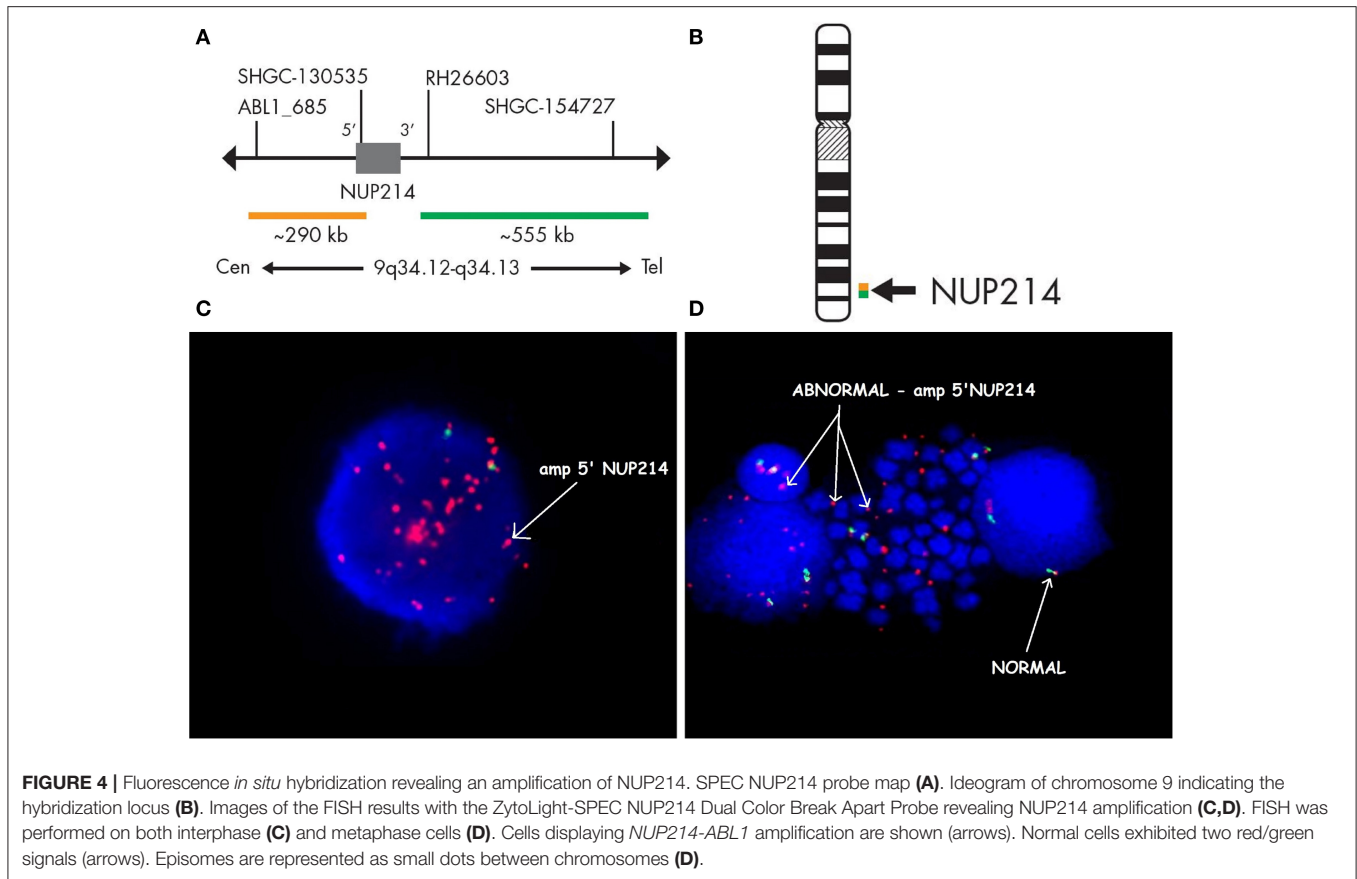
CNVs, two databases were applied: the UCSC database (<http://genome.ucsc.edu>) and Ensemble (<http://www.ensembl.org>). For microarray results, copy number polymorphisms were excluded based on comparison with the Database of Genomic Variants (<http://projects.tcag.ca/variation/>).

## Statistical Analysis

Clinical parameters, such as patient's age at diagnosis, white blood count (WBC), hemoglobin, and platelet count at diagnosis, percentage of blasts in peripheral blood and bone marrow, MRD level, and survival parameters, are presented as medians with interquartile ranges (IQRs) and were compared using Kruskal–Wallis analysis of variance or the  $\chi^2$  test. The association







observed in cytogenetic analysis as triplications between 11q13 and 11q12 of chromosome 11.

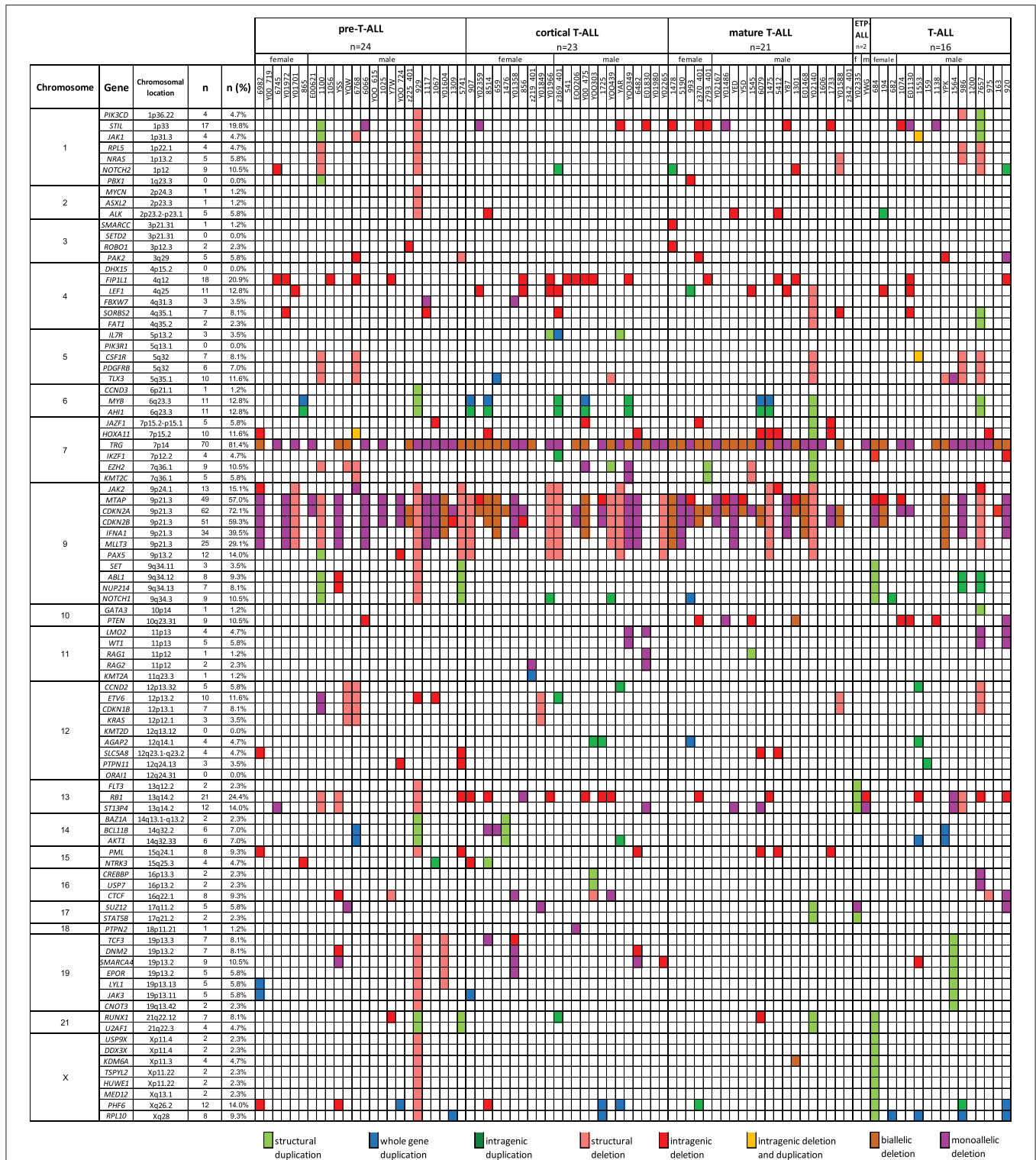
Case 986 presented an amplification at 9q34 with breakpoints in the NUP214 and ABL1 genes (Figure 3). All abnormalities appeared to be present in the entire leukemic population. Interphase FISH testing confirmed the amplification of NUP214 (Figure 4).

## The Landscape of CNAs in Childhood T-ALL

A total of 769 CNAs, including trisomies, duplications, deletions, and segmental loss of heterozygosity, were detected in 86 analyzed samples (Figure 5). The median number of CNAs per case was 16 (mean = 16.4, range = 1–49). Copy number abnormalities >10 Mb and <10 Mb accounted for 41% (10 trisomies, 75 duplications, 200 deletions, and 28 LOHs) and 59% (28 duplications of the whole gene, 42 intragenic duplications, 185 whole-gene monoallelic deletions, 85 whole-gene biallelic deletions, and 109 intragenic monoallelic deletions of gene and seven intragenic biallelic deletions), respectively.

We also selected a panel of 92 genes with documented roles in the pathogenesis of T-ALL and searched for CNAs within this gene panel. The results of the analysis are presented in Figure 6.

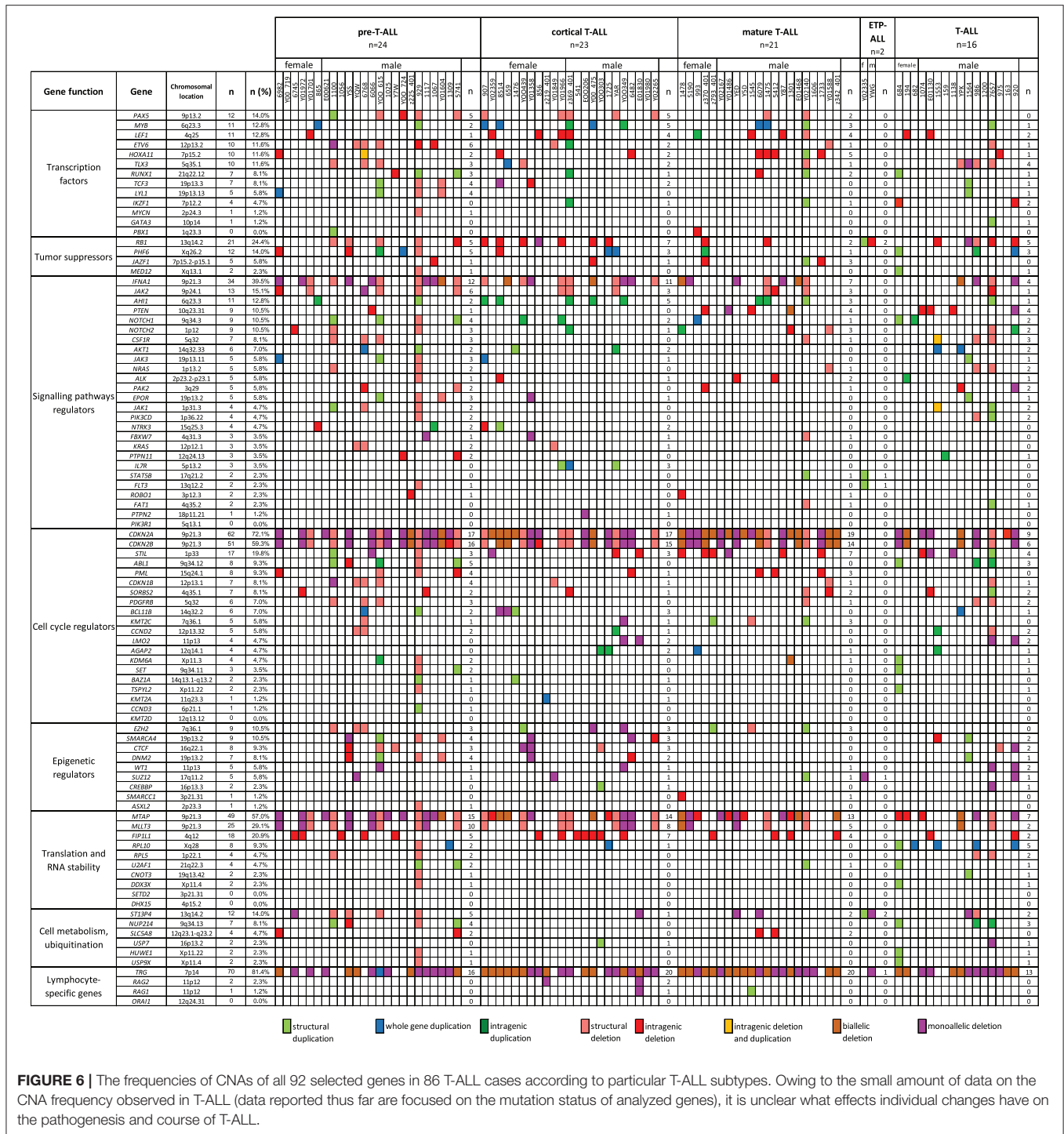
In our study, CNAs in *MYB* and *AIH1* (6q23.3) co-occurred in 11 cases (12.94%). We noticed whole-gene duplications associated with chromosomal alterations in 3 of 11 cases and intragenic duplications (from exon 3 to the 3' end) in 2 of 11 cases. In the remaining cases, we noticed coexisting whole-gene duplications in *MYB* and intragenic duplications in *AIH1* (from exon 27 to the 3' end). More importantly, CNAs in *MYB* and *AIH1* did not occur separately in any of our samples (Figure 5). In the case of *DNM2* (19p13.2), CNA occurred in seven cases (8.24%). In two cases, we found duplications associated with chromosomal alterations, and in five cases, we found deletions, including two cases with deletion associated with chromosomal alteration, one monoallelic deletion, and two with intragenic monoallelic deletions (from exon 2 to 3' end). Two cases of intragenic deletions in *ROBO1* (3p12.3) were found in this study. Three cases presented deletion of exon 2 in *PAK2* (3q29). In 18 samples (21.18%), intragenic deletions (from exon 6 to exon 10) in *FIP1L1* (4q12) were identified. The monoallelic intragenic deletion (from upstream region 1 exon to 3 exon) occurred in *LEF1* (4q25) in three cases. Monoallelic intragenic deletions covering the first exon were observed in *HOXA11-7p15.2* (nine cases), in *JAZF1-7p15.2-p15.1* (two cases), *PTEN-10q23.31* (two cases), and in *SLC5A8-12q23.1* (four cases). Monoallelic intragenic deletions of exons 3 and 4 were observed in six cases of *PML* (15q24.1).



**FIGURE 5 |** The frequencies of CNAs of all 92 selected genes in 86 T-ALL cases according to particular chromosomes.

Moreover, monoallelic intragenic deletions involving the region from exon 16 to exon 22 of gene *SORBS2* were found in five cases. Monoallelic intragenic deletions covering the last exon of

genes were found in *PTEN*-10q23.31 (from exon 2 to 3' end in three cases) and *STIL-1* (from exon 8 or 12 to 3' end in nine cases).

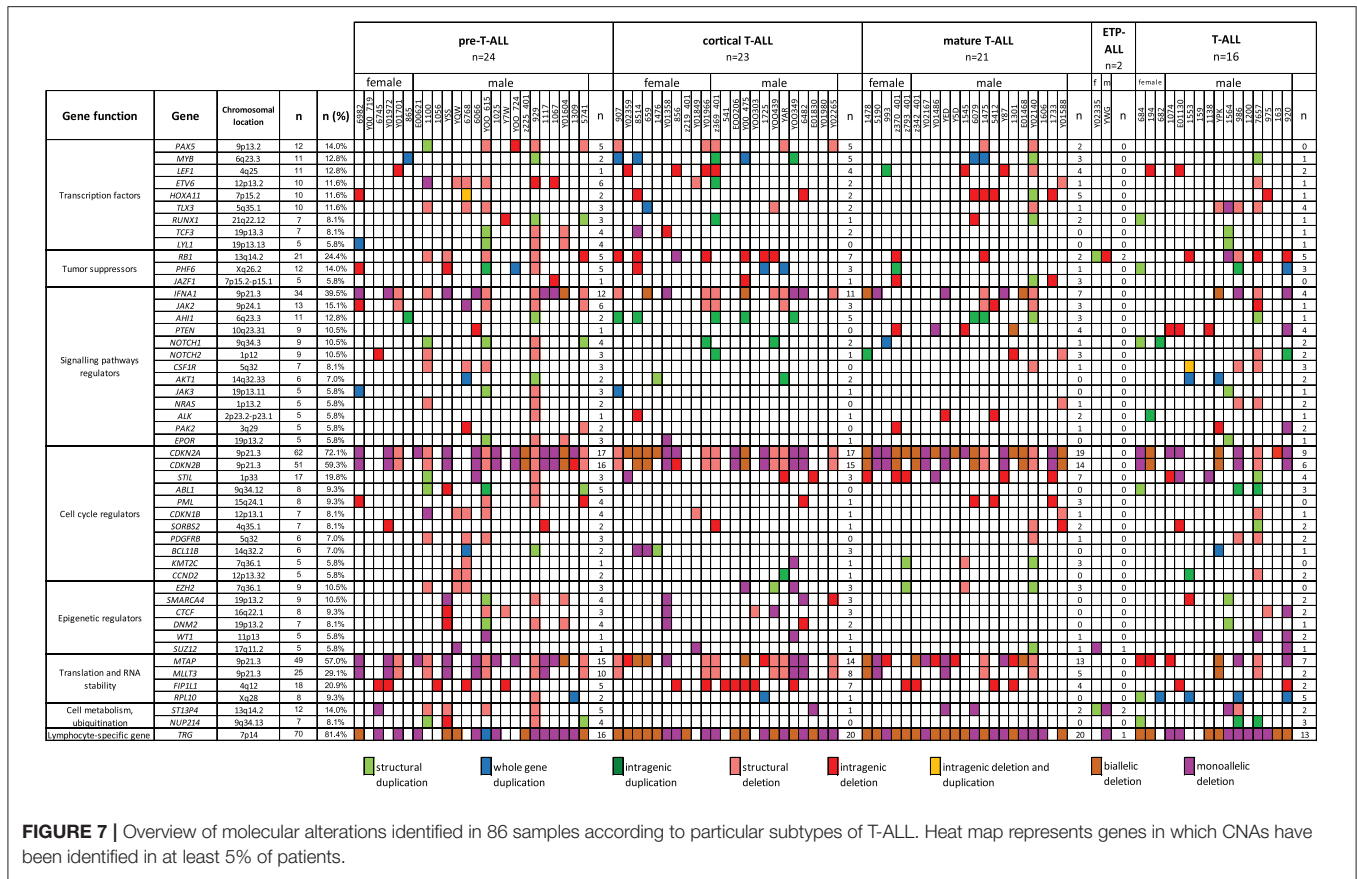


## Association of CNAs With Clinical Features and Response to Treatment

The relationship between microdeletions and microduplications of targeted genes (in which CNAs have been identified in at least 5% of patients), their functions, and immunophenotype is shown in **Figure 7**. Moreover, in some cases, multiple gene deletions

occurred in the same patients. The following interactions were still significant after correction for multiple testing: *CDKN2A-CDKN2B*, *CDKN2B-IFNA1*, *CDKN2B-MLLT3*, *PAX5-MLLT3*, *MTAP-CDKN2A*, *MTAP-CDKN2B*, *CDKN2A-IFNA1*, *IFNA1-MLLT3*, *MTAP-IFNA1*, *MTAP-MLLT3*, *PAX5-JAK2*, *JAK2-MLLT3*, *TRG-CDKN2A*, *PAX5-IFNA1*, *CDKN2A-MLLT3*,





*JAK2-IFNA1*, *PAX5-MTAP*, *PAX5-CDKN2B*, *TRG-MTAP*, *STIL-PTEN*, *TRG-CDKN2B*, *JAK2-ETV6*, *PAX5-RB1* (Figure 8). The frequency of CNAs was analyzed depending on the pathways and gene functions: transcriptional and epigenetic activity, regulation of the cell cycle, and translation or RNA stability (Figure 9). We found no significant predominance of alterations in any of the tested genes. The biggest difference was apparent for deletion of *CSF1R*, which was more prevalent in boys than in girls (13.6% vs. 0.0%), but this difference was not statistically significant ( $p = 0.05237$  without correction for multiple comparisons). We observed a significant difference in *NOTCH2* ( $p = 0.010$ ), *ALK* ( $p = 0.016$ ), *LEF1* ( $p = 0.023$ ), *TRG* ( $p = 0.001$ ), and *SLC5A8* ( $p = 0.023$ ) deletion frequencies in groups of patients with different WBC levels ( $p < 0.05$ ). In our cohort, we found a difference in *FIP1L1* ( $p = 0.004$ ), *LEF1* ( $p = 0.048$ ), *CSF1R* ( $p = 0.048$ ), *PDGFRB* ( $p = 0.025$ ), and *TLX3* ( $p = 0.035$ ) deletion frequency in a group of patients with poor response to steroids, compared to a group with good response ( $p < 0.05$ ). We noticed a significant difference in *FIP1L1* ( $p = 0.015$ ), *LEF1* ( $p = 0.007$ ), *MTAP* ( $p = 0.005$ ), and *CDKN2A* ( $p = 0.016$ ) deletion frequencies in the high-risk group of T-ALL patients in comparison to non-high-risk group ( $p < 0.05$ ). We also observed a difference in *FIP1L1* ( $p = 0.001$ ), *LEF1* ( $p = 0.044$ ), *PDGFRB* ( $p = 0.034$ ), *CDKN2A* ( $p = 0.027$ ), *CDKN2B* ( $p = 0.020$ ), and *STAT5B* ( $p = 0.008$ ) deletion frequency in group of patients with MRD flow cytometry (MRD FMC) levels on day 15 below 10%, compared to the group of patients with MRD FMC levels on day

15 exceeding 10% ( $p < 0.05$ ). We found a significant difference in *FIP1L1* ( $p = 0.003$ ), *WT1* ( $p = 0.014$ ), and *PHF6* ( $p = 0.043$ ) deletion frequency in the group of patients who relapsed in comparison to the group of patients who did not relapse ( $p < 0.05$ ). We observed a difference in *MTAP* ( $p = 0.003$ ), *IFNA1* ( $p = 0.007$ ), and *MLL3* ( $p = 0.008$ ) deletion frequency in the group of younger patients (<10 years old), compared to the group of older patients (>10 years old) ( $p < 0.05$ ) (Table 2). After adjusting for the multiple comparisons (92 genes), none of the associations between CNAs and clinical features analyzed was statistically significant.

In multivariate regression models, CNAs were predicted by following covariates: in *CDKN2A* by high risk (HR) ( $\beta = -1.86$ ,  $p = 0.030$ ) with HR decreasing the risk of CNAs; in *FIP1L1*, *PHF6*, and *WT1* by relapse (respectively,  $\beta = 1.67$ ,  $p = 0.046$ ;  $\beta = 2.15$ ,  $p = 0.032$ ; and  $\beta = 4.66$ ,  $p = 0.036$ ) with relapse increasing the risk of CNAs; in *TRG* by WBC ( $\beta = 0.001$ ,  $p = 0.007$ ) with increasing risk of CNAs with growing WBC; in *IFNA1* and *MLL3* by age (respectively,  $\beta = -0.12$ ,  $p = 0.039$ ; and  $\beta = -0.18$ ,  $p = 0.011$ ) with decreasing risk of CNAs with growing age (Table 3).

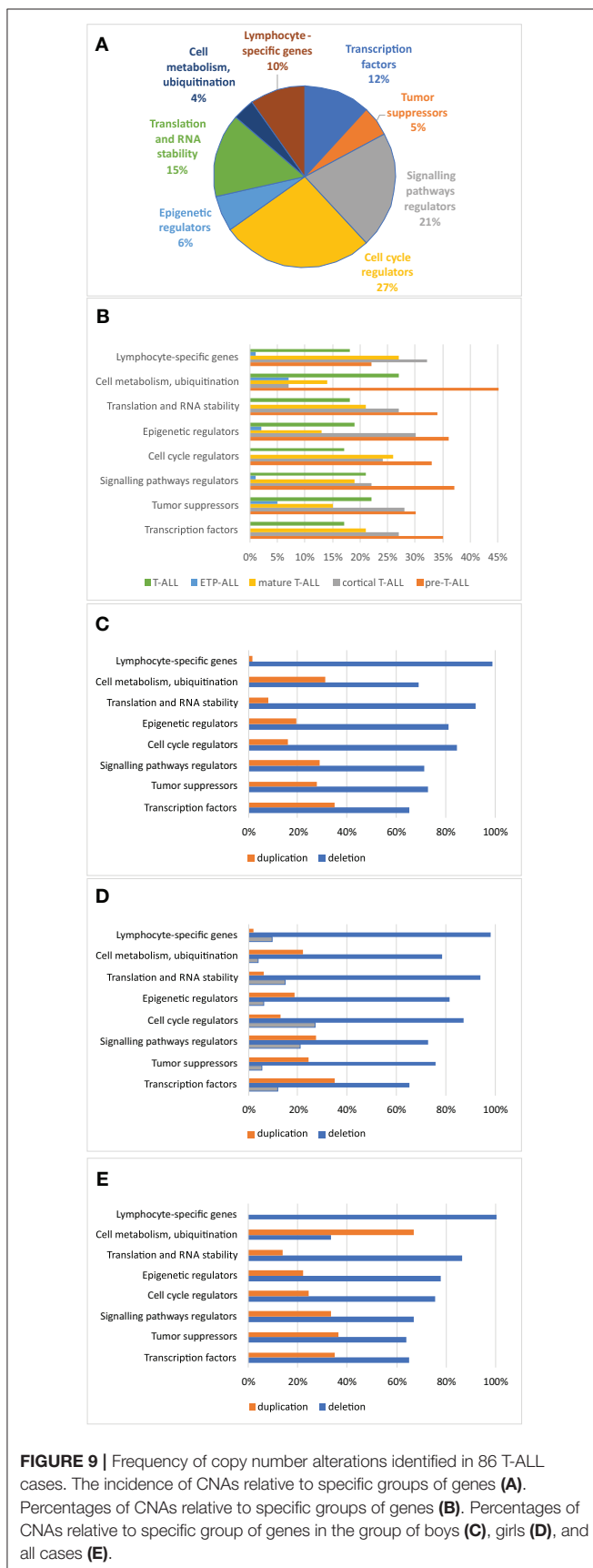
## DISCUSSION

In this study, we performed whole-genome characterization of CNAs in 86 pediatric patients diagnosed with T-ALL. T-cell

Genes	LEF1	PAX5	WT1	STIL	FIP1L1	TRG	JAK2	MTAP	CDKN2A	CDKN2B	TLX3	IFNA1	MLLT3	PTEN	ETV6	RB1	CTCF
LEF1	10	n=3 0.114	n=0 >0.999	n=3 0.369	n=3 0.430	n=10 0.199	n=4 0.040 >0.999	n=7 0.502	n=10 0.031 >0.999	n=7 0.510	n=1 >0.999	n=5 0.497	n=4 0.455	n=2 0.280	n=0 0.636	n=2 >0.999	n=0 0.588
PAX5		11	n=1 0.428	n=2 >0.999	n=1 0.445	n=9 >0.999	n=8 <0.001 0.001	n=11 0.132	n=11 0.029 >0.999	n=11 0.002 0.282	n=2 0.323	n=10 <0.001 0.028	n=10 <0.001	n=0 0.596	n=2 0.184	n=6 0.017 >0.999	n=1 >0.999
WT1			4	n=0 >0.999	n=2 0.192	n=3 >0.999	n=2 0.107	n=4 0.126	n=4 0.318	n=4 0.136	n=2 0.053	n=4 0.019 >0.999	n=3 0.064	n=1 0.363	n=2 0.087	n=3 0.038 >0.999	n=2 0.042 >0.999
STIL				15	n=2 0.727	n=14 0.284	n=2 >0.999	n=9 >0.999	n=13 0.537	n=8 0.382	n=0 0.349	n=5 0.388	n=4 0.542	n=6 0.001 0.088	n=1 0.084	n=2 0.504	n=0 0.341
FIP1L1					18	n=12 0.179	n=3 >0.999	n=7 0.118	n=11 0.383	n=8 0.282	n=1 0.678	n=5 0.416	n=3 0.376	n=3 0.388	n=1 >0.999	n=4 >0.999	n=3 0.356
TRG						69	n=11 >0.999	n=45 0.005 0.717	n=57 <0.001 0.011	n=46 0.012 >0.999	n=6 0.372	n=31 0.057	n=22 0.376	n=9 0.195	n=5 0.071	n=17 0.751	n=5 0.189
JAK2							13	n=11 0.033 >0.999	n=11 0.330	n=11 0.064	n=4 0.027 >0.999	n=11 <0.001 0.045	n=11 <0.001	n=0 0.344	n=4 0.014 >0.999	n=6 0.068	n=1 >0.999
MTAP								49	n=49 <0.001 <0.001	n=43 <0.001	n=7 0.288	n=34 <0.001	n=25 <0.001	n=5 >0.999	n=5 >0.999	n=15 0.071	n=5 >0.999
CDKN2A									62	n=51 <0.001 <0.001	n=7 >0.999	n=34 <0.001	n=25 <0.001	n=8 0.274	n=6 0.779	n=17 0.161	n=5 0.686
CDKN2B										51	n=7 0.293	n=34 <0.001	n=25 <0.001	n=3 0.156	n=6 >0.999	n=17 0.009 >0.999	n=5 >0.999
TLX3											9	n=7 0.024 >0.999	n=6 0.013 >0.999	n=0 0.588	n=4 0.006 0.788	n=6 0.004 0.576	n=2 0.196
IFNA1												34	n=25 <0.001	n=2 0.472	n=5 0.335	n=13 0.008 >0.999	n=5 0.251
MLLT3													25	n=1 0.434	n=4 0.260	n=10 0.021 >0.999	n=3 0.680
PTEN														9	n=0 0.634	n=2 >0.999	n=1 >0.999
ETV6															9	n=4 0.098	n=1 0.603
RB1																20	n=4 0.081
CTCF																	8

p<0.05       p<0.01

**FIGURE 8 |** Concomitance of gene deletions in 86 children with T-ALL. Yellow boxes indicate the total number of patients with a deletion. Other boxes indicate the number of patients with two specific abnormalities. Orange and blue colors indicate significant overlap associations in individual comparisons ( $p < 0.01$  or  $p < 0.05$ ), respectively. If association in individual comparison of two genes was significant,  $p$ -value corrected for multiple comparisons was added.



acute lymphoblastic leukemia can be classified based on both the immunophenotype and genetic alterations. Many of these variations are not mutually exclusive, and together, they lead to dysregulation of several signaling pathways involved in cell maturation, differentiation, proliferation, and apoptosis (8, 9, 11). The deletions/duplications rates in our cohort were similar to those previously described in the literature (9). One of the limitations of this study is lack of information about point mutations and gene fusions in our T-ALL samples, which would place specific CNAs that have been detected in the leukemic genome in the context of the remaining molecular defects.

The SNP array significantly improved the molecular characterization of pediatric ALL (10). Olsson et al. (13) presented 91 and 84% abnormal karyotypes after SNP array analysis of patients with B-cell precursor (BCP)-ALL and T-ALL, respectively. In our study, karyotypes were altered in 77% of pediatric T-ALL cases. Numerical aberrations were present in only 10 patients (single trisomies). In our T-ALL cohort, we found fewer structural abnormalities >10 Mb compared to BCP-ALL, no monosomies, and only segmental LOH. These observations correspond with those of other studies (14). Whole-chromosome Uniparental Disomies (UPIDs) are very rare (<1%) in T-ALL (14, 15), in contrast to pediatric BCP-ALL, where such LOHs are observed in 5% to 10% of cases (16). Moreover, Karrman et al. (15) suggested that deletions in T-ALL are more common [100/128 (78%)] than duplications [28/128 (22%)], which is consistent with our observations. However, repeated structural abnormalities are more common in T-ALL, as is LOH within chromosome 9. Additionally, the most frequent structural changes (>10 Mb) occurred on 9p21.3, as monoallelic or biallelic deletions within 9p21.3.

In two cases, SNP arrays allowed the identification of aberrations that were not detected by conventional karyotyping. The main advantage of SNP arrays is that DNA from tumor cells is used instead of mitotically dividing cells within the cell culture. In our cohort, *NUP214* and *ABL1* regions were amplified in 5 to 6% of T-ALL patients (17). We detected the *NUP214-ABL1* extrachromosomal (episomal) amplification of *ABL1* and *NUP214* and confirmed the findings by FISH. This study showed the impact of the SNP array in the detection of episomal amplification, which is undetectable using routine cytogenetic tests. The amplification of the *NUP214-ABL1* gene on episomes is associated with a higher risk of relapse and chemoresistance in T-ALL. However, aberrantly expressed Abl1 kinase due to *NUP214-ABL1* fusion is a target for therapy with Abl1 inhibitors. Therefore, identifying *ABL1* fusions in T-ALL is extremely important from a diagnostic and therapeutic point of view (18).

Moreover, we presented a patient (z219-401) with a chromothriptic-like pattern on chromosome 11. Chromothripsis has been associated with aggressive tumor progression and poor prognosis (19). In childhood, ALL chromothripsis has been described in rare BCP-ALL cases showing intrachromosomal amplification of chromosome 21(iAMP21) (20). The biological and clinical significance of this genome instability involving the other chromosomes in ALL needs further investigation (21).

**TABLE 2 |** A comparison between the presence of deletions in particular genes and clinical features of T-ALL patients.

(1) Gene	WBC			p-value	Adjusted p-value
	<5/L (n = 46)	5–10/L (n = 36)	> 10/L (n = 4)		
NOTCH2	9 (19.6)	0 (0.0)	0 (0.0)	0.010	0.939
ALK	2 (4.3)	1 (2.8)	2 (50.0)	0.016	>0.999
LEF1	2 (4.3)	8 (22.2)	1 (25.0)	0.023	>0.999
TRG	31 (67.4)	35 (97.2)	4 (100.0)	0.001	0.126
SLC5A8	0 (0.0)	3 (8.3)	1 (25.0)	0.023	>0.999
(2) Gene	Response to steroids		p-value	Adjusted p-value	
	Poor (n = 32)	Good (n = 54)			
FIP1L1	12 (37.5)	6 (11.1)	0.004	0.335	
LEF1	1 (3.1)	10 (18.5)	0.048	>0.999	
CSF1R	6 (18.8)	2 (3.7)	0.048	>0.999	
PDGFRB	5 (15.6)	1 (1.9)	0.025	>0.999	
TLX3	7 (21.9)	3 (5.6)	0.035	>0.999	
(3) Gene	Risk group		p-value	Adjusted p-value	
	HR (n = 43)	Non-HR (n = 43)			
FIP1L1	14 (32.6)	4 (9.3)	0.015	>0.999	
LEF1	1 (2.3)	10 (23.3)	0.007	0.663	
MTAP	20 (46.5)	29 (67.4)	0.050	>0.999	
CDKN2A	26 (60.5)	36 (83.7)	0.016	>0.999	
(4) Gene	MRD FMC 15 day		p-value	Adjusted p-value	
	<10% (n = 52)	> 10% (n = 34)			
FIP1L1	5 (9.6)	13 (38.2)	0.001	0.131	
LEF1	10 (19.2)	1 (2.9)	0.044	>0.999	
PDGFRB	1 (1.9)	5 (14.7)	0.034	>0.999	
CDKN2A	42 (80.8)	20 (58.8)	0.027	>0.999	
CDKN2B	36 (69.2)	15 (44.1)	0.020	>0.999	
STAT5B	0 (0.0)	2 (5.9)	0.008	0.735	
(5) Gene	Relapse		p-value	Adjusted p-value	
	No (n = 75)	Yes (n = 11)			
FIP1L1	12 (16.0)	6 (54.5)	0.003	0.307	
WT1	2 (2.7)	3 (27.3)	0.014	>0.999	
PHF6	8 (10.7)	4 (36.4)	0.043	>0.999	
(6) Gene	Age		p-value	Adjusted p-value	
	<10 years (n = 61)	> 10 years (n = 25)			
MTAP	41 (67.2)	8 (32.0)	0.003	0.235	
IFNA1	30 (49.2)	4 (16.0)	0.007	0.626	
MLLT3	23 (37.7)	2 (8.0)	0.008	0.720	

Data are presented as n (%),  $\chi^2$  test, or Fisher exact test. Only genes with  $p < 0.05$  are presented. Adjusted p-value for multiple comparisons: >0.999 for remaining genes.

WBC, white blood count; MRD FMC, minimal residual disease flow cytometry; HR, high-risk group; non-HR, non-high-risk group.

Data are presented as n (%),  $\chi^2$  test, or Fisher exact test. Only genes with  $p < 0.05$  are presented: (1) p-value adjusted for multiple comparisons (92 genes):  $p > 0.999$  for ALK, LEF1, SLC5A8,  $p = 0.939$  for NOTCH2,  $p = 0.126$  for TRG; (2) P-value adjusted for multiple comparisons (92 genes):  $p = 0.335$  for FIP1L1,  $p > 0.999$  for remaining genes; (3) p-value adjusted for multiple comparisons (92 genes):  $p = 0.663$  for LEF1,  $p > 0.999$  for remaining genes; (4) p-value adjusted for multiple comparisons (92 genes):  $p = 0.131$  for FIP1L1,  $p = 0.735$  for STAT5B, and  $p > 0.999$  for remaining genes; (5) p-value adjusted for multiple comparisons (92 genes):  $p = 0.307$  for FIP1L1,  $p > 0.999$  for remaining genes; (6) p-value adjusted for multiple comparisons (92 genes):  $p = 0.253$  for MTAP,  $p = 0.626$  for IFNA1,  $p = 0.720$  for MLLT3.

**TABLE 3** | Variables predictive of CNVs in particular genes in multivariate logistic regression coefficient- $\beta$  coefficient in logistic regression model with CNVs as the outcome variable.

Gene		Constant	Sex, male	WBC	Response to steroids, good	Risk, HR	MRD FMC 15 day	Relapse	Age, years
ALK	Coefficient	2.01	-0.25	0.001	-0.22	-17.4	-0.02	-15.33	-0.02
	$p$	>0.999	0.738	0.213	>0.999	0.997	0.836	0.997	0.862
CDKN2A	Coefficient	2.79	0.06	0.001	-0.64	-1.86	-0.01	2.07	-0.07
	$p$	<b>0.033*</b>	0.920	0.998	0.462	<b>0.030*</b>	0.705	0.068	0.274
CDKN2B	Coefficient	1.97	0.05	-0.001	-0.48	-0.58	-0.02	0.71	-0.08
	$p$	0.086	0.930	0.377	0.547	0.459	0.179	0.375	0.166
CSF1R	Coefficient	02.49	18.5	-0.001	-18.7	-18.6	0.02	1.01	0.09
	$p$	>0.999	0.995	0.309	0.997	0.997	0.296	0.413	0.435
FIP1L1	Coefficient	-1.41	-0.57	0.001	-1.20	0.57	-0.01	1.67	0.04
	$p$	0.328	0.407	0.526	0.217	0.603	0.759	<b>0.046*</b>	0.542
IFNA1	Coefficient	1.30	0.30	-0.001	-0.58	-0.30	-0.02	0.01	-0.12
	$p$	0.249	0.563	0.487	0.484	0.709	0.219	0.986	<b>0.039*</b>
LEF1	Coefficient	0.13	-1.41	0.001	-1.34	-1.88	0.06	2.46	-0.001
	$p$	0.994	0.072	0.495	0.994	0.992	0.145	0.178	0.989
MLL3	Coefficient	1.11	0.11	-0.001	-0.23	-0.47	-0.01	0.22	-0.18
	$p$	0.359	0.852	0.346	0.802	0.589	0.566	0.783	<b>0.011*</b>
MTAP	Coefficient	2.52	-0.06	-0.001	-0.58	-1.52	-0.01	1.69	-0.10
	$p$	<b>0.038*</b>	0.907	0.239	0.495	0.067	0.688	0.058	0.073
NOTCH2	Coefficient	15.72	0.19	-0.001	-16.81	-17.92	0.03	1.15	-0.02
	$p$	0.993	0.837	0.053	0.993	0.992	0.193	0.344	0.874
PDGFRB	Coefficient	-3.25	17.95	-0.001	-18.31	-17.55	0.02	0.23	0.07
	$p$	>0.999	0.996	0.584	0.997	0.997	0.307	0.872	0.577
PHF6	Coefficient	-0.68	-0.16	-0.001	-0.85	-1.49	-0.01	2.15	-0.01
	$p$	0.690	0.835	0.364	0.519	0.289	0.644	<b>0.032*</b>	0.920
SLC5A8	Coefficient	7.11	-0.09	0.001	-9.42	-18.83	0.13	-14.92	-0.16
	$p$	0.999	0.950	0.636	0.998	0.997	0.156	0.997	0.438
STAT5B	Coefficient	-0.07	-0.49	-0.004	0.03	-0.01	11.01	-0.42	17.77
	$p$	>0.999	>0.999	0.993	>0.999	0.996	0.993	>0.999	0.994
TLX3	Coefficient	13.69	0.36	0.001	-17.01	-15.78	0.01	-0.48	0.02
	$p$	0.994	0.698	0.492	0.993	0.994	0.592	0.701	0.827
TRG	Coefficient	1.12	0.08	0.001	0.45	-12.06	0.05	0.30	-0.13
	$p$	0.469	0.911	<b>0.007**</b>	0.646	0.255	0.143	0.778	0.115
WT1	Coefficient	-24.1	17.4	0.001	0.46	-1.16	0.06	4.66	0.12
	$p$	0.994	0.995	0.256	0.783	0.627	0.115	<b>0.036*</b>	0.511

\* $p < 0.05$ , \*\* $p < 0.01$ . The bold values are considered statistically significant.

Regarding abnormalities involving genes acting as transcription factors, *PAX5* was the most frequently altered gene [ $n = 12$  (14.0%)], which is in accordance with previous studies (11, 13). *PAX5* plays a key role in the development and maturation of B cells. Inactivation of *PAX5* as a result of deletion or mutation is characteristic of BCP-ALL. However, the loss of *PAX5* function may affect the pathogenesis of T-ALL because B-cell dedifferentiation to the progenitor cell stage with multilineage potential occurs (12, 22). A decrease in *PAX5* expression in T-ALL may also be associated with an increase in promoter methylation of this gene (23).

*RUNX1* genetic alterations are associated with poor prognosis because inactivation of this gene results in inhibition of T-cell transformation (24). The lack or reduced expression of *RUNX1* also affects the activity of *MYB*, *MYC*, and *GATA3* oncogenes,

which confirms the key role of *RUNX1* in the pathogenesis of T-ALL (24). In the described cohort of patients, *RUNX1*, *MYB*, and *GATA3* changes affected 8.1% ( $n = 7$ ), 12.8% ( $n = 11$ ), and 1.2% ( $n = 1$ ) of cases, respectively, which corresponds to the results that have been previously reported (4, 9, 25).

Deletion of the Wnt signaling central transcription mediator, *LEF1*, was found in 12.8% ( $n = 11$ ) of cases. The activity of this signaling pathway is closely related to the normal course of hematopoiesis; thus, the disturbed expression of key Wnt pathway molecules is characteristic of leukemias (26). Other studies reported a similar frequency of *LEF1* alterations in T-ALL (4, 27, 28). Moreover, Wang and Zhang (29) indicated that lower expression of *LEF1* is associated with ETP-ALL, but we cannot compare our data to published results because of the limited number of ETP-ALL cases ( $n = 2$ ) in our cohort.

The role of the tumor suppressor *PHF6* in the pathogenesis of T-ALL is not fully understood, but it remains one of the most frequently altered genes in this disease. *PHF6* alterations may be associated with glucocorticoid resistance (30). The number of *PHF6* CNAs in the analyzed group was similar to that previously reported (14% vs. 13%) (31). Moreover, Wendorff et al. (32) reported that the promotion of self-renewal capacity of hematopoietic stem cells driven by the loss of *PHF6* stimulates leukemia initiation T-ALL. Thus, *PHF6* deletion may be associated with leukemia development (32). Another negative regulator of the cell cycle is *RBI*; its alterations were detected in 24.4% ( $n = 21$ ) of cases. Another mechanism by which *RBI* activity might also be inhibited in T-ALL is by increased expression of mir-150 (33). We also observed alterations in *MED12* in 2.33% ( $n = 2$ ) of cases. Deletion or mutation within *MED12* oncogene may act as a tumor suppressor by inhibiting chemotherapy-induced apoptosis. This antiapoptotic effect by regulating the TGF $\beta$  pathway translates into an increase in resistance to anticancer drugs in the case of repression of *MED12* (34). We also identified CNAs within *JAZF1* in 5.81% of cases ( $n = 5$ ). Chromosomal aberrations involving this gene are associated with the pathogenesis of endometrial stromal tumors, and *JAZF1* itself acts as a transcription repressor (35). To date, no alterations within this gene have been described in the context of hematological diseases, including ALL.

Although *NOTCH1* alterations play a key role in the pathogenesis of T-ALL (36), *NOTCH1* abnormalities were detected in 10.5% ( $n = 9$ ) of the leukemic cases in our cohort. However, this difference may result from the fact that we only analyzed CNAs in the *NOTCH1* and *NOTCH2* genes. In T-ALL cases, mutations within the *NOTCH1* gene occurred much more frequently than *NOTCH1* deletions or duplications (36). Impaired *NOTCH1* activity may result in the activation of specific oncogenes (*c-MYC*, *NFkB*, or mTOR) or inhibition of tumor suppressor expression (*FBXW7*, *PTEN*, *CDKN1B*) (37). Therefore, we cannot estimate the impact of alterations within genes of *NOTCH1*-related signaling pathways, as we do not have data on *NOTCH1* mutations (37). Similarly, we observed that *FBXW7* alterations occurred much less frequently in our cohort compared to the frequency found in other studies (3.4% vs. 19–20%) (38, 39). *FBXW7* controls protein turnover by participating in the ubiquitination processes of these molecules and is also a part of the NOTCH signaling pathway. Therefore, the NOTCH signaling pathway is activated most often by genetic alterations within this gene or as a result of loss of function of its negative regulator, *FBXW7*. The results from previous studies do not clearly indicate the prognostic value of *NOTCH1* and *FBXW7* lesions in T-ALL (38–41). Nevertheless, SNP arrays are not appropriate for *NOTCH1* and *FBXW7* mutation status assessment as a single technique, and other sequencing methods need to be used to obtain a complete profile of aberrations in these genes in T-ALL patients.

Activation of Notch1 pathway can also occur as a result of dysregulation within other signaling pathways, such as the PI3K/mTOR pathway. We identified genetic alterations involving the PI3K/mTOR pathway genes in 20 cases (*PTEN*:  $n = 10$ , *AKT1*:  $n = 6$ , *PIK3CD*:  $n = 4$ ), which corresponded to the data

previously described (42–44). *PTEN* is a negative regulator of Notch1 pathway, and disturbances in its activity (due to mutation or deletion) may be associated with unfavorable long-term outcomes in some cases of T-ALL (45, 46). Additionally, a change in *PTEN* expression secondary to *NOTCH1*-activating mutation may result in cellular resistance to  $\gamma$ -secretase inhibitors (43). A few studies have shown that *PTEN* alterations can affect the risk of relapse; therefore, screening of alterations of this gene could potentially improve risk group stratification (43, 44).

In the study cohort, other signaling pathways, such as the JAK/STAT pathway and RAS pathway, may be affected by CNAs within particular genes. The CNA rates of genes of these pathways are similar to data previously described (31.4% vs. 25% and 11.6% vs. 15%, respectively) (9). Impaired activity of *RAS* occurs as a result of *NRAS* and *KRAS* deletion or *FLT3* duplication. For each of these genes, the alteration rate was similar to that previously described in childhood lymphoblastic leukemia (9).

The interferon  $\alpha 1$  gene, *IFNA1*, had a mutation frequency of 39.5% ( $n = 34$ ). Interferons have antiproliferative activity, and their expression influences cancer initiation or progression (47).

We observed an increased frequency of alterations within the genes involved in cell cycle regulation. *CDKN2A/B* deletions [ $n = 62$  (72.1%) and  $n = 51$  (59.3%), respectively] showed even higher prevalence than that described in other studies. *CDKN2A/B* acts as a tumor suppressor (48, 49). According to our results, only *CDKN2A* deletions were associated with a specific mature immunophenotype of T-ALL (90.5%), in comparison with cortical-T-ALL (73.9%) and pre-T-ALL (68.0%) ( $p = 0.04$ ). Our data also showed a statistically significant co-occurrence of deletions in *CDKN2A* and *LEF*, *PAX5*, *TRG*, and *MTAP*, as well as *CDKN2B* and *PAX5*, *TRG*, *JAK2*, and *MTAP*. Co-occurrence of *CDKN2A* and *CDKN2B* deletions was reported in 51 cases. Some studies estimated the effect of deletion (mono- or biallelic) of *CDKN2A/B* on outcome, but the results are inconclusive (50). Similar concomitance between particular gene deletions (*WT1*, *IFNA*, *RBI*, and *CTCF*) was previously reported by Zhang et al. (51). We did not observe the co-occurrence between *SUZ12* and *CDKN2AB* alterations, although Noronha et al. (4) indicate such concomitance in their study. The study of Karrman et al. (15) revealed that *CDKN2A* deletions were significantly associated with a high WBC count, but we did not find a similar association in our cohort. Interestingly, the co-occurrence of *CDKN2A* deletions with *CDKN2B*, *MTAP*, or *MLLT3* deletions in the same T-ALL patients was previously reported by Yeh et al. (52). Moreover, our results indicate similar concomitance between *MTAP* and *CDKNAB* deletions. However, in contrast to our results, Yeh et al. (52) reported no significant correlation between the status of CNAs and clinical features, such as sex, age, and WBC.

Deletions and duplications within the genes responsible for posttranscriptional processing play an important role in leukemogenesis. Copy number abnormality rate in this group of genes is comparable to the data described in the literature (9). In our cohort, the most frequently mutated gene in this group was *FIP1L1* [ $n = 18$  (20.9%)]. *FIP1L1* status was significantly related to the classical prognostic factors in T-ALL. Alterations in *FIP1L1* were more often associated with

a worse response to prednisone and a higher MRD level on day 15 (>10%). To date, no case indicating a relationship between intragenic *FIP1L1* deletion and T-ALL pathogenesis has been described. However, in myeloproliferative disorders with eosinophilia, *FIP1L1-PDGFRBA* fusion has been observed. The *FIP1L1-PDGFRBA* protein is composed of the first 12 exons of *FIP1L1* and from truncated exon 12 (containing the last 17 amino acids) to exon 23 of *PDGFRA*. The *FIP1L1-PDGFRBA* fusion protein is a constitutively activated tyrosine kinase that joins the first 233 amino acids of *FIP1L1* to the last 523 amino acids of *PDGFRA* (53). The presence of such a rearrangement allows the implementation of targeted treatment with the tyrosine kinase inhibitor imatinib (54, 55). In our cohort, the intragenic deletion of *FIP1L1* (exons 5–8 or exons 6–10) was found. To date, there are no published data indicating the role of the *FIP1L1* intragenic deletion in T-ALL pathogenesis.

Inactivation of epigenetic factors results in arresting T-cell development, which was confirmed by studies on ETP-ALL mouse models (56). Furthermore, *WT1* haploinsufficiency may promote the accumulation of somatic mutations or mutations of other epigenetic factors in progenitor cells, leading to tumor transformation (57). Recurrent microduplications in T-ALL are less common. *MYB* (6q23.3) is gained in ~10% of cases as a focal duplication involving solely the *MYB* locus (12). Interestingly, in our study, CNAs in *MYB* exist only in association with *AIH1* alterations, accounting for 12.94% of cases.

## CONCLUSIONS

Our analysis shows that the profile of CNAs is associated with the immunophenotypic and clinical features of T-ALL. For the first time, we described the link between *FIP1L1* deletions and the clinical features of patients with T-ALL. Moreover, we observed an increased percentage of cases with concomitant alterations within the *MYB* and *AIH1* genes. We also characterized structural aberrations >10 Mbp, which was lower in our cohort than in BCP-ALL, and we showed that SNP array definitely improved the molecular characterization of pediatric T-ALL. In some specific cases including those harboring *NUP214-ABL1*, SNP array might be valuable

diagnostic tool for selecting patients who may benefit from targeted therapy.

## DATA AVAILABILITY STATEMENT

The datasets presented in this study can be found in online repositories. The names of the repository/repositories and accession number(s) can be found below: NCBI GEO under the accession GSE147381.

## ETHICS STATEMENT

The studies involving human participants were reviewed and approved by Ethics Committee of Medical University of Silesia. Written informed consent to participate in this study was provided by the participants' legal guardian/next of kin.

## AUTHOR CONTRIBUTIONS

ML planned the study. ML, AP, and WM were responsible for the conception and design of the study. JZ, ŁS, KS, OH, and TS shared patients' clinical data. BS and JT conducted the laboratory work. ML and MW were responsible for the interpretation of genetic data. MB was responsible for additional statistical analysis. ML, MW, and JZ were responsible for the acquisition of literature for the manuscript. ML and MW wrote the paper. AP, WM, JK, and TS supervised the paper. All authors have read and approved the final manuscript.

## FUNDING

This work was supported by the National Centre for Research and Development, Poland (grant number: STRATEGMED3/304586/5/NCBR/2017). AP was supported by Polpharma Scientific Foundation (grant number: 9/XV/2016).

## SUPPLEMENTARY MATERIAL

The Supplementary Material for this article can be found online at: <https://www.frontiersin.org/articles/10.3389/fonc.2020.01184/full#supplementary-material>

## REFERENCES

- Muffly L, Larson RA. Improving outcomes in childhood T-cell acute lymphoblastic leukemia: promising results from the Children's Oncology Group incorporating nelarabine into front-line therapy. *Transl Pediatr.* (2012) 1:120–2. doi: 10.3978/j.issn.2224-4336.2012.09.03
- Chiaretti S, Zini G, Bassan R. Diagnosis and subclassification of acute lymphoblastic leukemia. *Mediterr J Hematol Infect Dis.* (2014) 6:e2014073. doi: 10.4084/mjhid.2014.073
- Coustan-Smith E, Mullighan CG, Onciu M, Behm FG, Raimondi SC, Pei D, et al. Early T-cell precursor leukaemia: a subtype of very high-risk acute lymphoblastic leukaemia. *Lancet Oncol.* (2009) 10:147–56. doi: 10.1016/S1470-2045(08)70314-0
- Noronha EP, Marques LVC, Andrade FG, Thuler LCS, Terra-Granado E, Pombo-de-Oliveira MS, et al. The profile of immunophenotype and genotype aberrations in subsets of pediatric T-cell acute lymphoblastic leukemia. *Front Oncol.* (2019) 9:316. doi: 10.3389/fonc.2019.00316
- Pui CH, Robison LL, Look AT. Acute lymphoblastic leukaemia. *Lancet.* (2008) 371:1030–43. doi: 10.1016/S0140-6736(08)60457-2
- Lustosa de Sousa DW, de Almeida Ferreira FV, Cavalcante Félix FH, de Oliveira Lopes MV. Acute lymphoblastic leukemia in children and adolescents: prognostic factors and analysis of survival. *Rev Bras Hematol Hemoter.* (2015) 37:223–9. doi: 10.1016/j.bjhh.2015.03.009
- Hunger SP, Mullighan CG. Acute lymphoblastic leukemia in children. *N Engl J Med.* (2015) 373:1541–52. doi: 10.1056/NEJMra1400972
- Mullighan CG. The molecular genetic makeup of acute lymphoblastic leukemia. *Hematol Am Soc Hematol Educ Prog.* (2012) 2012:389–96. doi: 10.1182/asheducation.V2012.1.389.3798360
- Liu Y, Easton J, Shao Y, Maciaszek J, Wang Z, Wilkinson MR, et al. The genomic landscape of pediatric and young adult T-lineage acute

- lymphoblastic leukemia. *Nat Genet.* (2017) 49:1211–18. doi: 10.1038/ng.3909
10. Lejman M, Zawitkowska J, Styka B, Babicz M, Winnicka D, Zaucha-Prazmo A, et al. Microarray testing as an efficient tool to redefine hyperdiploid paediatric B-cell precursor acute lymphoblastic leukaemia patients. *Leuk Res.* (2019) 83:1–11. doi: 10.1016/j.leukres.2019.05.013
  11. Thakral D, Kaur G, Gupta R, Benard-Slagter A, Savola S, Kumar J, et al. Rapid identification of key copy number alterations in B- and T-cell acute lymphoblastic leukemia by digital multiplex ligation-dependent probe amplification. *Front Oncol.* (2019) 9:871. doi: 10.3389/fonc.2019.00871
  12. Karrman K, Johansson B. Pediatric T-cell acute lymphoblastic leukemia. *Genes Chromosomes Cancer.* (2017) 56:89–116. doi: 10.1002/gcc.22416
  13. Olsson L, Lundin-Ström KB, Castor A, Behrendtz M, Biloglav A, Norén-Nyström U, et al. Improved cytogenetic characterization and risk stratification of pediatric acute lymphoblastic leukemia using single nucleotide polymorphism array analysis: a single center experience of 296 cases. *Genes Chromosomes Cancer.* (2018) 57:604–7. doi: 10.1002/gcc.22664
  14. Kawamata N, Ogawa S, Zimmermann M, Kato M, Sanada M, Yamamoto G, et al. Molecular allelotyping of pediatric acute lymphoblastic leukemias by high-resolution single nucleotide polymorphism oligonucleotide genomic microarray. *Blood.* (2008) 111:776–84. doi: 10.1182/blood-2007-05-088310
  15. Karrman K, Castor A, Behrendtz M, Forestier E, Olsson L, Ehinger M, et al. Deep sequencing and SNP array analyses of pediatric T-cell acute lymphoblastic leukemia reveal NOTCH1 mutations in minor subclones and a high incidence of uniparental isodisomies affecting CDKN2A. *J Hematol Oncol.* (2015) 8:42. doi: 10.1186/s13045-015-0138-0
  16. Lundin KB, Olsson L, Safavi S, Biloglav A, Paulsson K, Johansson B. Patterns and frequencies of acquired and constitutional uniparental isodisomies in pediatric and adult B-cell precursor acute lymphoblastic leukemia. *Genes Chromosomes Cancer.* (2016) 55:472–9. doi: 10.1002/gcc.22349
  17. Papanhausen P, Kelly CA, Zhang X, Tepperberg J, Burnside RD, Schwartz S. Multidisciplinary analysis of pediatric T-ALL: 9q34 gene fusions. *Cancer Genet.* (2019) 231–2:1–13. doi: 10.1016/j.cancergen.2018.12.002
  18. Zhou MH, Yang QM. NUP214 fusion genes in acute leukemia. *Oncol Lett.* (2014) 8:959–62. doi: 10.3892/ol.2014.2263
  19. Ratnaparkhe M, Hlevnjak M, Kolb T, Jauch A, Maass KK, Devens F, et al. Genomic profiling of acute lymphoblastic leukemia in ataxia telangiectasia patients reveals tight link between ATM mutations and chromothripsis. *Leukemia.* (2017) 31:2048–56. doi: 10.1038/leu.2017.55
  20. Li Y, Schwab C, Ryan S, Papaemmanuil E, Robinson HM, Jacobs P, et al. Constitutional and somatic rearrangement of chromosome 21 in acute lymphoblastic leukaemia. *Nature.* (2014) 508:98–102. doi: 10.1016/S0959-8049(14)50362-0
  21. Koltsova AS, Pendina AA, Efimova OA, Chiryaeva OG, Kuznetsova TV, Baranov VS. On the complexity of mechanisms and consequences of chromothripsis: an update. *Front Genet.* (2019) 10:393. doi: 10.3389/fgene.2019.00393
  22. Cobaleda C, Jochum W, Busslinger M. Conversion of mature B cells into T cells by dedifferentiation to uncommitted progenitors. *Nature.* (2007) 449:473–7. doi: 10.1038/nature06159
  23. Hütter G, Kaiser M, Neumann M, Mossner M, Nowak D, Baldus CD, et al. Epigenetic regulation of PAX5 expression in acute T-cell lymphoblastic leukemia. *Leuk Res.* (2011) 35:614–19. doi: 10.1016/j.leukres.2010.11.015
  24. Choi A, Illendula A, Pulikkan JA, Roderick JE, Tesell J, Yu J, et al. RUNX1 is required for oncogenic Myb and Myc enhancer activity in T-cell acute lymphoblastic leukemia. *Blood.* (2017) 130:1722–33. doi: 10.1182/blood-2017-03-775536
  25. Chen B, Jiang L, Zhong ML, Li JF, Li BS, Peng LJ, et al. Identification of fusion genes and characterization of transcriptome features in T-cell acute lymphoblastic leukemia. *Proc Natl Acad Sci USA.* (2018) 115:373–8. doi: 10.1073/pnas.1717125115
  26. Kirstetter P, Anderson K, Porse BT, Jacobsen SE, Nerlov C. Activation of the canonical Wnt pathway leads to loss of hematopoietic stem cell repopulation and multilineage differentiation block. *Nat. Immunol.* (2006) 7:1048–56. doi: 10.1038/ni1381
  27. Montaña A, Forero-Castro M, Marchena-Mendoza D, Benito R, Hernández-Rivas JM. New challenges in targeting signaling pathways in acute lymphoblastic leukemia by NGS approaches: an update. *Cancers.* (2018) 10:110. doi: 10.3390/cancers10040110
  28. Gutierrez A, Sanda T, Ma W, Zhang J, Grebliunaite R, Dahlberg S, et al. Inactivation of LEF1 in T-cell acute lymphoblastic leukemia. *Blood.* (2010) 115:2845–51. doi: 10.1182/blood-2009-07-234377
  29. Wang M, Zhang C. Low LEF1 expression is a biomarker of early T-cell precursor, an aggressive subtype of T-cell lymphoblastic leukemia. *PLoS ONE.* (2020) 15:e0232520. doi: 10.1371/journal.pone.0232520
  30. Xiang J, Wang G, Xia T, Chen Z. The depletion of PHF6 decreases the drug sensitivity of T-cell acute lymphoblastic leukemia to prednisolone. *Biomed Pharmacother.* (2019) 109:2210–17. doi: 10.1016/j.biopha.2018.11.083
  31. Teachey DT, Pui CH. Comparative features and outcomes between pediatric T-cell and B-cell acute lymphoblastic leukaemia. *Lancet.* (2019) 20:142–54. doi: 10.1016/S1470-2045(19)30031-2
  32. Wendorff AA, Quinn SA, Rashkovan M, Madubata CJ, Ambesi-Impimbato A, Litzow MR, et al. Phf6 loss enhances HSC self-renewal driving tumor initiation and leukemia stem cell activity in T-ALL. *Cancer Discov.* (2019) 9:436–51. doi: 10.1158/2159-8290.CD-18-1005
  33. Miao MH, Ji XQ, Zhang H, Xu J, Zhu H, Shao XJ. miR-590 promotes cell proliferation and invasion in T-cell acute lymphoblastic leukaemia by inhibiting RB1. *Oncotarget.* (2016) 7:39527–34. doi: 10.18632/oncotarget.8414
  34. Huang S, Hölzel M, Knijnenburg T, Schlicker A, Roepman P, McDermott U, et al. MED12 controls the response to multiple cancer drugs through regulation of TGF- $\beta$  receptor signaling. *Cell.* (2012) 151:937–50. doi: 10.1016/j.cell.2012.10.035
  35. Tsuyoshi H, Yoshida Y. Molecular biomarkers for uterine leiomyosarcoma and endometrial stromal sarcoma. *Cancer Sci.* (2018) 109:1743–52. doi: 10.1111/cas.13613
  36. Weng AP, Ferrando AA, Lee W, Morris JP, Silverman LB, Sanchez-Irizarry C, et al. Activating mutations of NOTCH1 in human T cell acute lymphoblastic leukemia. *Science.* (2004) 306:269–71. doi: 10.1126/science.1102160
  37. Breit S, Stanulla M, Flohr T, Schrappe M, Ludwig WD, Tolle G, et al. Activating NOTCH1 mutations predict favorable early treatment response and longtermoutcome in childhood precursor T-cell lymphoblastic leukemia. *Blood.* (2006) 108:1151–7. doi: 10.1182/blood-2005-12-4956
  38. Mansur MB, Hassan R, Barbosa TC, Splendore A, Jotta PY, Yunes JA, et al. Impact of complex NOTCH1 mutations on survival in paediatric T-cell leukaemia. *BMC Cancer.* (2012) 12:9. doi: 10.1186/1471-2407-12-9
  39. Yuan L, Lu L, Yang Y, Sun H, Chen X, Huang Y, et al. Genetic mutational profiling analysis of T cell acute lymphoblastic leukemia reveal mutant FBXW7 as a prognostic indicator for inferior survival. *Ann Hematol.* (2015) 94:1817–28. doi: 10.1007/s00277-015-2474-0
  40. Natarajan V, Bandapalli OR, Rajkumar T, Sagar TG, Karunakaran N. NOTCH1 and FBXW7 mutations favor better outcome in pediatric South Indian T-cell acute lymphoblastic leukemia. *J Pediatr Hematol Oncol.* (2015) 37:e23–30. doi: 10.1097/MPH.0000000000000290
  41. Asnafi V, Buzyn A, Le Noir S, Baleyrier F, Simon A, Beldjord K, et al. NOTCH1/FBXW7 mutation identifies a large subgroup with favorable outcome in adult T-cell acute lymphoblastic leukemia (T-ALL): a Group for Research on Adult Acute Lymphoblastic Leukemia (GRAALL) study. *Blood.* (2009) 113:3918–24. doi: 10.1182/blood-2008-10-184069
  42. Zuurbier L, Petricoin EF III, Vuerhard MJ, Calvert V, Kooi C, Buijs-Gladdines JG, et al. The significance of PTEN and AKT aberrations in pediatric T-cell acute lymphoblastic leukemia. *Haematologica.* (2012) 97:1405–13. doi: 10.3324/haematol.2011.059030
  43. Paganin M, Grillo MF, Silvestri D, Scapinello G, Buldini B, Cazzaniga G, et al. The presence of mutated and deleted PTEN is associated with an increased risk of relapse in childhood T cell acute lymphoblastic leukaemia treated with AIEOP-BFM ALL protocols. *Br J Haematol.* (2018) 182:705–11. doi: 10.1111/bjh.15449
  44. Szarzynska-Zawadzka B, Kunz JB, Sedek L, Kosmalska M, Zdon K, Biecek P, et al. PTEN abnormalities predict poor outcome in children with T-cell acute lymphoblastic leukemia treated according to ALL IC-BFM protocols. *Am J Hematol.* (2019) 94:E93–6. doi: 10.1002/ajh.25396
  45. Tesio M, Trinquand A, Ballerini P, Hypolite G, Lhermitte L, Petit A, et al. Age-related clinical and biological features of PTEN abnormalities in T-cell acute lymphoblastic leukaemia. *Leukemia.* (2017) 31:2594–600. doi: 10.1038/leu.2017.157



46. Vicente C, Schwab C, Broux M, Geerdens E, Degryse S, Demeyer S, et al. Targeted sequencing identifies associations between IL7R-JAK mutations and epigenetic modulators in T-cell acute lymphoblastic leukemia. *Haematologica*. (2015)100:1301–10. doi: 10.3324/haematol.2015.130179
47. Khouri R, Silva-Santos G, Dierckx T, Menezes SM, Decanine D, Theys K, et al. A genetic IFN/STAT1/FAS axis determines CD4 T stem cell memory levels and apoptosis in healthy controls and Adult T-cell Leukemia patients. *Oncoimmunology*. (2018) 7:e1426423. doi: 10.1080/2162402X.2018.1426423
48. Jang W, Park J, Kwon A, Choi H, Kim J, Lee GD, et al. CDKN2B downregulation and other genetic characteristics in T-acute lymphoblastic leukemia. *Exp Mol Med*. (2019) 51:4. doi: 10.1038/s12276-018-0195-x
49. Genesà E, Lazarenkov A, Morgades M, Berbis G, Ruiz-Xiville N, Gomez-Marzo P, et al. Frequency and clinical impact of CDKN2A/ARF/CDKN2B gene deletions as assessed by in-depth genetic analyses in adult T cell acute lymphoblastic leukemia. *J Hematol Oncol*. (2018)11:96. doi: 10.1186/s13045-018-0639-8
50. Agarwal M, Bakhshi S, Dwivedi SN, Kabra M, Shukla R, Seth R. Cyclin dependent kinase inhibitor 2A/B gene deletions are markers of poor prognosis in Indian children with acute lymphoblastic leukemia. *Pediatr Blood Cancer*. (2018) 65:e27001. doi: 10.1002/pbc.27001
51. Zhang L, Ding L, Holmfeldt L, Wu G, Heatley SL, Payne-Turner D, et al. The genetic basis of early T-cell precursor acute lymphoblastic leukaemia. *Nature*. (2012) 481:157–63. doi: 10.1038/nature10725
52. Yeh TC, Liang DC, Liu HC, Jaing TH, Chen SH, Hou JY, et al. Clinical and biological relevance of genetic alterations in pediatric T-cell acute lymphoblastic leukemia in Taiwan. *Pediatr Blood Cancer*. (2019) 66:e27496. doi: 10.1002/pbc.27496
53. Gotlib J, Cools J. Five years since the discovery of FIP1L1-PDGFR $\alpha$ : what we have learned about the fusion and other molecularly defined eosinophilias. *Leukemia*. (2008)22:1999–2010. doi: 10.1038/leu.2008.287
54. Reiter A, Gotlib J. Myeloid neoplasms with eosinophilia. *Blood*. (2016) 129:704–14. doi: 10.1182/blood-2016-10-695973
55. Olsson-Arvidsson L, Norberg A, Sjögren H, Johansson B. Frequent false-negative FIP1L1-PDGFR $\alpha$  FISH analyses of bone marrow samples from clonal eosinophilia at diagnosis. *Br J Haematol*. (2019) 188:e76–9. doi: 10.1111/bjh.16340
56. Wang C, Oshima M, Sato D, Matsui H, Kubota S, Kazumasa A, et al. Ezh2 loss propagates hypermethylation at T cell differentiation-regulating genes to promote leukemic transformation. *J Clin Invest*. (2018) 128:3872–86. doi: 10.1172/JCI94645
57. Pronier E, Bowman RL, Ahn J, Glass J, Kandoth C, Merlinsky TR, et al. Genetic and epigenetic evolution as a contributor to WT1-mutant leukemogenesis. *Blood*. (2018)132:1265–78. doi: 10.1182/blood-2018-03-837468

**Conflict of Interest:** The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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