Complex genotype-phenotype relationships shape the response to treatment of Down Syndrome Childhood Acute Lymphoblastic Leukaemia

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Abstract

Extensive genetic and epigenetic variegation has been demonstrated in many malignancies. Importantly, their interplay has the potential to contribute to disease progression and treatment resistance. To shed light on the complex relationships between these different sources of intra-tumour heterogeneity, we explored their relative contributions to the evolutionary dynamics of Acute Lymphoblastic Leukaemia (ALL) in children with Down syndrome, which has particularly poor prognosis. We quantified the tumour propagating

potential of genetically distinct sub-clones using serial transplantation assays and SNP-

arrays. While most leukaemias were characterized by a single dominant subclone, others

were highly heterogeneous. Importantly, we provide clear and direct evidence that

genotypes and phenotypes with functional relevance to leukemic progression and treatment

resistance can co-segregate within the disease. Hence, individual genetic lesions can be

restricted to well-defined cell immunophenotypes, corresponding to different stages of the

leukemic differentiation hierarchy and varied proliferation potentials. As a result of this

difference in fitness, which can be accurately quantified via competitive transplantation

assays, matching diagnostic, post-treatment, and relapse leukaemias can be dominated by

different genotypes, including pre-leukemic clones persisting throughout the disease

progression and treatment. Intriguingly, plasticity also appears to be a temporally defined

property that can segregate with genotype. These results suggest that Down Syndrome ALL

should be viewed as a complex matrix of cells exhibiting genetic and epigenetic

heterogeneity that foster extensive clonal evolution and competition. Therapeutic

intervention reshapes this 'ecosystem' and may provide the right conditions for the

preferential expansion of selected compartments and subsequent relapse.

Introduction:

The existence of intra-tumour heterogeneity, both genetic and epigenetic, has been

documented in both solid and haematological malignancies[1–3]. In studies of childhood acute

lymphoblastic leukaemia (ALL), we and others previously used multicolour-FISH and whole

genome sequencing to show that these tumours are clonally variegated, and to demonstrate

through serial transplantation assays, that leukaemia tumour propagating cells (TPCs) are

also genetically heterogeneous[4,5]. This finding was supported by our later genetic analysis of

immunophenotypically distinct ALL diagnostic subpopulations which demonstrated similar

distribution of subclones within a disease-specific primitive B-cell compartment (stemB,

$34^+38^-19^+$) and the disease bulk in most but, crucially, not all leukaemias[6]. Such experiments

relied on high-depth bulk whole-genome sequencing followed by single-cell qPCR and,

together with our earlier work, suggested that genotype and phenotypes related to

differentiation stage and cell-cycle state might not necessarily co-segregate within the

disease. Importantly, as chemotherapy-induced selection acts on these same phenotypic

traits[6–8], this lack of segregation also underpinned the survival of most genotypes following

the early stages of childhood ALL treatment (namely, induction chemotherapy – where

maximum cytotoxicity is achieved); as we demonstrated using a xenograft model of

transplantation and treatment of primary leukaemias[6].

Large-scale comparisons of matching diagnostic and relapse samples from paediatric ALL

patients, have nonetheless previously reported that in some patients a subset of diagnostic

subclones, or of their later evolutionary descendants, can be selectively enriched at the time

of disease recurrence[9,10]. Assuming that these clones did not become dominant as a result

of stochastic selection, this finding opens up the possibility that fitness-based genotypic

selection might potentially occur on a longer timescale than that of induction chemotherapy.

It is not unreasonable to speculate that in a treatment protocol that lasts up to four years

even subtle differences in fitness might eventually lead to clonal dominance. Another

Another possibility is that in certain settings, genotype and treatment-associated phenotypes

do in fact co-segregate (either already at diagnosis or later because of further genomic

evolution). In support of this latter hypothesis, our analysis of $34^{+}38^{-}19^{+}$ (StemB) BCP-ALL

cells indicated that in a small proportion of leukaemias (1 out of 5 analysed), specific

genotypes might indeed be preferentially enriched within developmentally more primitive and

quiescent disease compartments[6]. Furthermore, recent evidence suggests that mutations

induced by chemotherapy treatment itself (particularly those targeting p53) might be

responsible for evolution of new and fitter subclones, inducing drug-resistance and leading to

relapse[11].

To shed light on the complex genotype-phenotype relationship that shape ALL, we here

conduct a comprehensive investigation into the relative contribution of different sources of

intratumor heterogeneity, namely genetic makeup and epigenetically determined properties -

particularly cell cycle state, TPC activity and differentiation stage - to disease initiation,

progression, and treatment resistance. In doing so we focused our attention on Down's

syndrome associated ALL as this rarer disease subtype has so far remained poorly

characterised while being associated with considerably worse prognosis than ALL in children

without Down syndrome[12,13].

Altogether our results support the idea that ALL should be viewed as a complex matrix of

cells exhibiting genetic and epigenetic heterogeneity. Our work provides evidence for, in

some patients, diversification in functional properties amongst immunophenotypically-distinct

leukemic subpopulations. In theis setting, those rare leukemic cells which remain after

induction chemotherapy, and primarily account for minimal residual disease (MRD), are

highly enriched for the stemB phenotype; suggesting that the more primitive and less

proliferative leukemic compartments likely constitute the pool of cells from which relapse

arises. Crucially, in this setting where genotype and resistance-associated phenotypes co-

segregate, bottleneck selection at phenotype level is accompanied by evidence of a clonal

sweep.

Methods:

Patient samples

Diagnostic and follow-up ALL bone marrow (BM) samples were obtained after informed

consent and the approval of the relevant research ethics committee from patients at the

Paediatrics Haematology Unit, Lund University Hospital, Sweden. Samples from 4 children

with Down's ALL were analyzed. For further patient characteristics see Table 1.

Cell separation, flow cytometric analysis and cell sorting

Total mononuclear cells (MNCs) were isolated from BM samples by ficoll gradient centrifugation and directly cryopreserved in DMSO for later use. In some cases, CD34$^+$ cells were enriched by magnetic bead separation (StemCell Technologies or Miltenyi). CD34-enriched cells or MNCs were stained with anti-human CD19 PE (BD-Pharmingen), CD34 FITC (BD-Pharmingen) and CD38 APC (BD-Pharmingen). Cells were resuspended in staining buffer containing 10%FBS, DAPI at 0·1μg/ml and flow cytometric analysis and cell sorting was perfomed using a BD FACSAriaII cell sorter using FACSDiva Software v6.0 (BD Biosciences). Cell subsets were defined as I.) HSC (34$^+$38$^-$19$^-$) II.) Stem/B (34$^+$38$^-$19$^+$) III.) Progenitor (34$^+$38$^+$19$^-$) IV.) ProB (34$^+$38$^+$19$^+$) and V.) PreB/Mature B-cells (34$^-$19$^+$). Further data analysis was performed using FlowJo v10.7 (Tree star) software.

Cell cycle analysis of leukaemia subpopulations

Cells were stained with anti-himan CD19-PECy5 (BioLegend), then fixed and permeabilized with 1.6% paraformaldehyde and 90% ice-cold methanol. This was followed by staining with anti-human CD34-APC (BD), CD38-PETxR (Invitrogen) and Ki67-FITC (Becton-Dickinson). For DNA content analysis cells were incubated with 0·5μg/mL DAPI. Stained cells were analyzed on a FACS LSRII SORP using FACSDiva Software v6.0 (BD Biosciences).

FISH analysis

Cells were fixed on slides in methanol/acetic acid fixative (3:1 vol/vol) and then hybridized with BCR-ABL1 FISH probes using the LSI BCR/ABL1 Dual Color Dual Fusion Translocation probes (Abbot). The slides were analyzed with an Olympus BX51 microscope equipped with epifluorescence and a triple band pass filter. Images were captured by using a Sensys charge-coupled device camera (Photometrics, Tucson, AZ) and MacProbe software (Applied Imaging, Newcastle upon Tyne, U.K.). When possible, and unless otherwise specified at least 100-150 nuclei were analyzed in each sample.

SNP-Array

SNP-Arrays we performed as previously describe in Schuh et al., 2012[14].

NSG-mouse transplantation assay

Primary childhood BCP-ALL cells were transplanted into 8-12 weeks old NOD/SCID IL2Rγ[null] (NSG) sub-lethally irradiated mice (either females or males) via intramedullary injection. To minimize possible adverse effects of sublethal irradiation, mice were administered acid water for a week prior to the procedure, and Baytril (resuspended at 25.5 mg/kg in the drinking water) for the 2 weeks following it. Sub-lethal irradiation was achieved with a single dose of 375 cGy. Unless otherwise stated, each mouse received 2×10^5 primary leukaemia cells resuspended in 40μl PBS 0.5% FBS. In the case of secondary limiting dilution assays, a specified equal dose of treated and control leukemic cells harvested from the BM of primary recipients was injected. Twelve-week post-injection mice were sampled by BM aspiration and the percentage of human engraftment was evaluated by flow cytometry ($hCD45/(hCD45^+ + mCD45^+)$). At the same time, human cells were also FACS sorted for downstream applications. Mice displaying at least 70% human engraftment were then randomly assigned to either control or treatment groups (see below for details on the treatment protocol). At the end of the treatment window all mice were culled, and their tibias, femurs, pelvises, spleen, and brain harvested, and cells were then stained for flow cytometry analysis (see above).

Modeling of clonal evolution

To reconstruct the clonal composition of the samples, we developed an algorithm similar to Stiehl at al., 2021[15], which is tailored to the specific data generated in this study. The algorithm systematically generates all possible combinations of different clones and checks for each combination whether it can explain the experimental data. This is done by expressing the allele frequencies in the data as a linear combination of the clones present in the sample. Whenever there exists a linear combination with positive coefficients summing up to 100%, the considered combination of clones is in line with the data. Details are provided in the supplementary materials.

Small cell numbers RNA sequencing

Equivalent cell numbers (400 cells per sample) were flow sorted directly into 800µl Trizol reagent (Invitrogen) and snap frozen in dry ice (long term storage at -80C). At the time of extraction, the samples were thawed at RT and 160ul of chloroform was added to each. Following a centrifugation step the RNA was isolated from the aqueous phase and precipitated through the addition of equal volumes of isopropanol supplemented with 20µg linear polyacrylamide. Samples were washed twice in 80% ethanol (first wash over night at 4°C, second wash 5 minutes at RT). RNA pellets were resuspended in 3-15µl of diethylpyrocarbonate treated water (DEPC). RNA was then quantified by loading of 0.5-1ul on an Agilent Bionalyser RNA 6,000 pico chip. Where possible equivalent amounts of total RNA (100pg) from all samples were used for first strand synthesis with the SmartERv3 kit (Takara Clontech) followed by 15-18 cycles of amplification (according to manufacturers' instruction). cDNA was purified on Agencourt AMPureXP magnetic beads, washed twice with fresh 80% ethanol and eluted in 17µl elution buffer. 1µl cDNA was quantified with Qubit dsDNA HS (Molecular Probes) and checked on an Agilent Bioanalyser high sensitivity DNA chip. Sequencing libraries were produced from 150pg input cDNA using Illumina Nextera XT library preparation kit. A 1:4 miniaturized version of the protocol was adopted (see "Fluidigm Single-Cell cDNA Libraries for mRNA sequencing", PN_100-7168_L1). Tagmentation time was 5mins, followed by 12 cycles of amplification using Illumina XT 24 or 96 index primer kit. Libraries were then pooled (1-2ul per sample depending on the total number of samples) and purified with equal volumes (1:1) of Agencourt AMPureXP magnetic beads. Final elution was in 66-144ul of resuspension buffer (depending on the total number of pooled samples). Libraries were checked on an Agilent Bioanalyser high sensitivity DNA chip (size range 150-2000bp) and quantified by Qubit dsDNA HS (Molecular Probes). Libraries were sequenced on Illumina® NextSeq 500 using 150bp paired end kits as per manufacturer's instructions.

Bioinformatics

Bulk RNAseq data were processed with the nf-core/rnaseq pipeline (v1.1-blic; https://github.com/UCL-BLIC/rnaseq). Reads were initially trimmed with TrimGalore! v0.5.0 (https://github.com/FelixKrueger/TrimGalore) and aligned with STAR v2.7.0d[16]. Gene expression was quantified with featureCount v1.6.3[17]. Quality was assessed with FastQC v0.11.8 (http://www.bioinformatics.babraham.ac.uk/projects/fastqc/) and RSeQC v2.6.4[18] and summarised with MultiQC v1.6[19]. The gene expression matrix was analysed with DESeq2 v1.26.0[20] for data transformation, outlier detection, normalization, PCA and differential gene expression analyses. All analyses use vst (variance stabilizing transformation) transformed data. A PCA analysis with the top 500 most variable genes highlighted an outlying sample that was removed from downstream analysis. Gene expression signatures for B-cell lineages[21], BCR signaling, and B-cell differentiation[22] were mapped on the PCA by calculating the arithmetic mean of the PC contributions from the gene in the signatures. Gene Set Enrichment Analysis (GSEA) was performed with the fgsea v1.12.0 R package[23] using the following signatures: Hallmark from MSigDB[24], B-cell lineage[21], cell-cycle state[25] and BCR signalling[22].

Data availability

SNP-array and RNAseq data are available at EGA under accession number EGAS00001005968.

Results

We here explored the role of intra-tumour heterogeneity in the progression and response to treatment of Down's Syndrome-associated ALL. We initially focused on genetic variegation and described the phylogenetic hierarchies of diagnostic material from 4 patients. Serial transplantation of bulk ALL samples into immune-deficient NSG-mice was concomitantly utilized to evaluate the tumour propagating potential and dynamic behaviours of the different

subclones identified. SNP-arrays were performed on diagnostic specimens as well as cells

isolated from the BM of primary and secondary transplanted mice (see schemata in Figure

1A). In patients 1, 3 and 4, a single dominant clone was detected both at diagnosis and in all

primary and secondary transplanted mice (Figure 1B). In patient 2, instead, a more complex

picture was observed, with several subclones coexisting within the tumour and contributing

to a different extent to engraftment across recipients. In this patient we detected two

deletions on chromosomes 9 and 14 respectively, as well as an amplification on

chromosome X (Figure 1C). Considering the mutational mosaicism, we identified in some of

the engrafted mice, these observations allowed reconstruction of a simple clonal

phylogenetic architecture (Figure 1D, and Supp Material for Mosaicism data). However, a

higher-depth genetic analysis of chromosome 9 suggested a far more complicated clonal

landscape. A heterozygous chromosome 9 deletion was detected across engrafted animals

with different degrees of homozygous involvement: to the extreme of copy neutral loss of

heterozygosity (cnLOH) (Figure 2A). A simple explanation to this observation is that pre-

existing sub-clones present at diagnosis at sub-threshold levels, were expanded through

secondary transplantation until reaching detection levels. Alternatively, continuous

broadening of the chromosome 9 deletion could have occurred over the short time frame of

the serial transplantation assays. This latter hypothesis would be indicative of active ongoing

clonal evolution. Based on the mosaicism data, and assuming no new mutations developed

during transplantation, we were able to reconstruct a more comprehensive phylogenetic tree.

To do so, we developed a custom-made computational model which we used to calculate all

the theoretically possible diagnostic clone-combinations supported by our SNPs data

analysis (Supplementary Table 1). In total 15-28 diagnostic clones were necessary to

explain the mutational read-out from the xenografts. Of note, since SNP-arrays are limited in

both genomic resolution (i.e., deletion size) and sensitivity to low abundance copy number

variants (CNV) it is possible that our model might underestimate the number of clones

(Figure 2B and supplementary methods). Nonetheless, this finding clearly illustrates the high

clonal complexity of ALL and is in line with our earlier single-cell WGS based

characterization of genetic heterogeneity in non-Down Syndrome ALL, which also identified

branching clonal architectures with up to 19 descendant subclones[6].

We subsequently sought to shed light on the phenotype-genotype relationship that defines

the properties of individual leukemic subclones, by performing SNP analysis on bulk tumour

cells and flow-sorted immunophenotypically-distinct ALL subpopulations from two Down's

syndrome ALL patients (Patient 2 and 3), both before and after xenotransplantation (Figure

3A and Supplementary Figure 1). In this analysis we detected identical clonal architectures

within diagnostic bulk, stemB and proB cells isolated from each patient leukaemia (Figure

3BC). All samples analysed from patient 3 confirmed the existence of a single dominant

leukemic clone (Figure 3C and supplementary table 1). In addition to the founding

chromosome 21 amplification, the third patient (Patient 4) we examined harboured a BCR-

ABL translocation. FISH analysis showed that in this leukaemia the fusion gene marked

proB- and preB-like sub-compartments, but not cells with the more primitive and disease-

specific stemB ($34^+38^-19^+$) phenotype[26] (Figure 4A). Surprisingly, in serial transplantation

experiments flow-sorted stemB cells only gave rise to grafts with stemB like cells but not

more differentiated (proB-preB) B-cell phenotypes. These stemB like grafts remained BCR-

ABL-negative. Similarly, proB cells derived grafts only contained proB and preB cells, all of

which retained the BCR-ABL-fusion. Collectively these data suggest lack of phenotypic

plasticity amongst diagnostic cells from this patient.

We next compared the relative fitness of the two subpopulations (stemB and proB) in a

competitive transplantation setting facilitated by (i) the presence of the BCR-ABL

translocation as a clonotypic marker and (ii) the non-interchangeability of their

immunophenotypes. We transplanted NSG-mice with different ratios of stemB and proB ALL

cells and read out engraftment. In all cases the BCR-ABL$^+$ proB cells displayed increased

engraftment fitness, most clearly exemplified in the setting where proB cells represented

only 10% of the initial inoculum and yet constituted 95% of the resultant leukaemia (Figure

4B). Reasoning that enhanced fitness may stem from a proliferative advantage we next

analysed the cell cycle status of engrafted cells. In all engrafted recipients proB cells

displayed a more active cycling status compared to their stemB counterparts, which had a

significantly larger proportion of dormant (G_0) cells (Figure 4C and Supplementary Figure 2).

We subsequently speculated that the different cycling properties of these compartments

could be expected to affect their relative sensitivity to therapy. To experimentally determine

how these different ALL sub-compartments behaved during therapy, we analysed leukemic

subpopulations from serially collected diagnosis, remission, and relapse samples. ProB-like

ALL cells were indeed preferentially eradicated during chemotherapy with only stemB ALL

cells remaining robustly detectable during remission (Figure 5A). After induction

chemotherapy extremely few proB-like cells remained, none of which carried the BCR-ABL

translocation. While the behaviour of these subpopulations during treatment contrasts starkly

with their fitness as assessed by xenotransplantation, it is entirely consistent with the known

relative vulnerability of cycling versus dormant cells to chemotherapy. Similarly, to what

previously observed in the context of bacteria and antibiotics resistance, these data also

support the idea of the existence of a trade-off on phenotypic traits between environments

with and without drug[27]. The reduced competitive ability of the less proliferative species

(stemB cells) in the absence of treatment hence becomes an advantage in the presence of

chemotherapy, while the highly proliferative proB cells pay a cost to their initial fitness.

Interestingly, the relapse sample was populated by both stemB and proB-like cells (Figure

5A). However, in contrast to the situation at diagnosis leukemic proB cells present at relapse

were BCR-ABL-negative (data not shown) and both relapse stemB and proB cells showed

phenotypic plasticity upon serial transplantation (Figure 5B). This data is somewhat

counterintuitive in that it argues against the simplest origins of relapse from the predominant

stemB or proB clones present at diagnosis. The stemB compartment at presentation lacked

phenotypic plasticity in that it was unable to regenerate proB- or preB-like cells. Since the

proB cells at relapse were BCR-ABL-negative, relapse could also not have been seeded by

the predominant diagnostic proB-like compartment that was positive for BCR-ABL.

Taking advantage of the xenograft system further investigated the phenotypic properties of

diagnostic and relapse cells at the molecular level by performing small cell number bulk RNA

sequencing (RNAseq) on cells harvested from either primary and or secondary xenografts.

We hence characterised six distinct cell populations i) diagnostic stemB cells, ii) diagnostic

proB cells iii) relapse stemB cells harvested from secondary xenografts injected with stemB

cells isolated from primary xenografts iv) relapse proB cells harvested from secondary

xenografts injected with stemB cells isolated from primary xenografts, v) relapse proB cells

harvested from secondary xenografts injected with proB cells isolated from primary

xenografts, and vi) relapse stemB cells harvested from secondary xenografts injected with

proB cells isolated from primary xenografts. Principal Component Analysis (PCA) showed

clustering of samples based on disease stage. Interestingly, diagnostic cells also clustered

separately based on differentiation state, while all relapse populations clustered closely

together (Figure 5C). This finding resonates well with our observation that in this patient

leukaemia, relapse but not diagnostic cells displayed high cell plasticity. While the PC1

component of the PC captured most of the sample variance (47.15), PC2 clearly separated

diagnostic proB cells and all relapse subpopulations from diagnostic stemB cells. Pathways

analysis on the genes driving this PC component revealed enrichment for TNF-alpha via NF-

KB signalling cascade (Supplementary Figure 3C), which has previously been implicated in

regulation of cell survival, proliferation and stemness[28].

Gene set enrichment analysis (GSEA) using publicly available gene expression

signatures[2921] also validated our flow cytometry analysis results in that compared to

diagnostic proB cells and relapse cells, diagnostic stemB cells downregulated cell

proliferation associated genes and expressed higher levels of markers typical of more

immature stages of the haematopoietic and B-cell differentiation hierarchy (such as HSC and

MLP cells) (Figure 5C-D and Supplementary Figure 3A-B). Most B-lineage cells from ALL

patients are thought to be arrested at some point of the transition from the proB to preB

stage, therefore broadly allowing the classification of ALL leukaemias into two distinct

subtypes based on preBCR function. Those arrested at the preBI stage rely upon

IL7R/STAT5 signaling (preBCR$^-$), while those arrested at preBII stage signal by the preBCR

(preBCR$^+$)[22]. The frequency of pre-BCR$^+$ or pre-BCR$^-$ cases has also been suggested to

associate with different cytogenetic subtypes of childhood ALL. For example, the majority of

BCR-ABL1$^+$ leukaemias tend to display constitutive active cytokine signalling via activation

of STAT5 and repression of BCL6. Herein, however, we found that these two genetically and

phenotypically distinct disease subsets can coexist within the same leukaemia. Interestingly,

contrary to expectations, diagnostic proB cells, which carried the BCR-ABL translocation and

expressed markers associated with later stages of the B-cell differentiation (Figure 5C),

displayed stronger expression of the preBCR receptor components and signalling cascade,

while the BCR-ABL$^-$ diagnostic stemB cells preferentially upregulated various genes involved

in the STAT5 signalling pathway (Supplementary Figure 3D).

We subsequently used SNP arrays to genetically characterise different phenotypic

compartments at presentation, relapse and after serial transplantation and track the

phylogenetic origins of the relapse propagating clones (Figure 6). SNP analysis of the

relapse samples indicated that it lacked several of the distinctive genetic abnormalities which

characterised either the dominant stemB compartment at diagnosis or the dominant proB.

However, the relapse hierarchy shared with the diagnostic compartments deletions on both

chromosomes 12 and 14. Of note, diagnosis and relapse clones also shared a common VDJ

rearrangement (data not shown).

Altogether, this data implies the survival over treatment of a common evolutionary ancestor

not detectable at diagnosis which gave rise to the relapse hierarchy through further

diversification. Accordingly, analysis of a bulk remission sample identified mosaic deletion of

chromosome 14 in the absence of del12 and any other obvious genetic abnormality.

Therefore, suggesting that this mosaic deletion is not a germline event and, likely, the

earliest genomic alteration (also preceding del12). To understand the basis of the mosaicism

observed in the remission sample, we separated CD3+ and CD3- cells and performed SNP

analysis. The CD3- compartment contained no evidence of the del14. In contrast the CD3+

cells showed a complex pattern of chromosome 14 deletions indicative of a progressive

multi-step evolutionary process (Supplementary table 2). We next traced these del14

variants back to the diagnostic and relapse stemB and proB compartments, as well as the

CD3+ compartment isolated at diagnosis. This spectrum of 14 deletions provides a branched

framework for subsequent evolution of stemB and proB clones seen at diagnosis and

relapse. Thus, using chromosomal deletions as markers of lineal ancestry we have here

reconstructed the evolutionary history of a Down's ALL from the predisposing event (Trisomy

21), through the acquisition of initiating lesions giving rise to a preleukemic hierarchy, into

further transformation marked by the onset of frank leukaemia. We also monitored the

response of this complex leukemic hierarchy to chemotherapy through the analysis of

matching remission and relapse specimens. Beyond the genetic considerations our work

emphasizes the role of cellular state primarily at the level of dormancy and cell cycle

indicating vulnerability to chemotherapy.

Discussion

Many studies have explored the extent and impact of different, individually assessed,

sources of intratumor heterogeneity on the evolution of tumours (particularly under

treatment). Our work, however, argues that cancer should be viewed as a complex matrix of

genetic and epigenetic heterogeneity such as cellular phenotype and cell cycle properties

which, altogether rather than in isolation, provide a substrate for tumour evolution during

disease progression in response to the selective pressure exercised by therapeutic

interventions.

The idea that epigenetically determined cell states, contribute to functional heterogeneity

amongst tumour cells by controlling fundamental cellular properties such as, for example,

cell identity, phenotype and cell-cycle stage is paradigmatically exemplified by the cancer

stem cell (CSC) hypothesis; which postulates that cells within a tumour are organized in a

hierarchical fashion reflecting lineage relationships and tumorigenic potential, and that the

maintenance of cancer clones is uniquely dependent on the most primitive CSCs with self-

renewal capacity[30–32]. This notion, which was pioneered in the context of blood malignancies,

remains however extensively debated. While phenotypic heterogeneity - primarily at the level

of immune-phenotype – has been widely observed in ALL, disparate data exist concerning

the existence of leukaemia stem cells (LSCs) within this disease. Earlier work by Rehe et al.,

showed that many cellular fractions contain LSC-potential at a similar frequency and that the

majority of - if not all - cells may be able to propagate the disease[33]. Importantly, the same

study also found evidence for phenotypic plasticity within different subpopulations upon

serial transplantation in immune-deficient mice. Rare LSCs have nonetheless been

conclusively identified in acute myeloid leukaemia (AML), where functional evidence for their

existence has been provided[34–37]. However, even in this context, their exact phenotype

remains elusive, with more recent studies uncovering LSC activity across different cellular

subsets[38]. We here provide evidence that although different Down's ALL disease

compartment can all contribute to leukaemia propagation *in vivo,* these same compartments

display largely distinct fitness; both in the presence and in the absence of chemotherapy. In

this scenario, more primitive stem-like disease compartments are outcompeted by highly

proliferative early-B cell progenitors in the absence of treatment. However, this selective

advantage is reversed in the presence of chemotherapy, which selects for quiescent cells.

While the genetically heterogenous nature of childhood ALL has been demonstrated by

multiple earlier studies, our current work provides the first direct evidence for genetic

diversity of cancer propagating cells in patients with Down's syndrome ALL. Our SNP array

analysis of serially transplanted ALLs shows that leukemic cells undergo a process of

gradual branching clonal evolution demonstrated by the coexistence within a leukaemia of

multiple subclones bearing alterations of the same genomic locus characterised by distinct

breakpoints.

Our earlier work showed that in many ALL leukaemias genotype and phenotypes with

relevance to chemoresistance do not segregate. As a result, induction chemotherapy does

not select at the level of genotype. We here identified and studied the evolutionary history of

an unusual case of BCR-ABL+ Down's ALL in which genotype and phenotype do instead

clearly segregate. In this patient's leukaemia the stemB cells were fusion-gene-negative

while proB/pre-B cells were fusion-gene-positive. This case allowed us to reconstruct in

detail the complex genotype-phenotype relationship characterizing different leukemic cell

fractions labelled by both distinct genetic makeup and fitness. We hence observed that

highly quiescent and developmentally more primitive cells preferentially escaped treatment.

As these cells carried only a subset of diagnostic genetic alterations, the relapse disease

was genetically divergent from the diagnostic specimen. Reconstruction of the leukaemia

phylogenetic tree demonstrated that the dominant population at relapse originated from a

rare pre-leukemic clone. This was likely present at diagnosis below detection level and, in

line with its likely low proliferative activity, did not extensively expand upon transplantation.

We first demonstrated the existence of pre-leukemic cells in ALL in studies of mono-

chorionic twins of whom one was diagnosed with TEL-AML1-positive ALL[26,39]. Blood spot

studies from Guthrie-cards further showed that these pre-leukemic cells can precede overt

leukaemia by up to 12 years and are also detectable in patients that never go on to overt

disease[40,41]. A case report of a 7-year old boy diagnosed with BCP-ALL with a mixed

lympho-myeloid phenotype previously suggested that BCR-ABL+ non-Down leukaemic cells

could reactivate to drive disease recurrence more than two decades from initial diagnosis[42].

Our new data add to our understanding of this elusive population by showing that Down

Syndrome BCP-ALL chr21+ cells, which have not yet acquired the BCR-ABL translocation

labelling the more proliferative proB compartment disease, can preferentially survive

chemotherapy and, therefore, represent a reservoir for relapse initiation. Whether these cells

are quiescent principally as a function of their unique developmental origin[43] or rather of their

genotype[44] or a combination of both remains to be investigated and would be important to

understand if new therapeutic interventions targeting these cells are to be designed. This

notwithstanding, pre-leukemic clones, even when rare, should clearly be considered part of

the fully transformed leukaemia landscape and may upon selection by therapy provide an

additional cellular source for subsequent relapse.

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Authorship Contributions

CL and VT performed experiments; RC analysed the SNP array; JH analysed the RNAseq

data with input from VT; AS interpreted and supervised SNP-array data analyses; TS

developed the computational model; CL and PSW performed the cell cycle analysis, VB

carried out FISH analyses; AC provided patient data and samples; SAC, HF and PSW

performed cell sorting. CL, VT, SEWJ and TE designed the study and interpreted data. CL,

VT and TE wrote the paper with critical input from other authors.

Conflict of Interest Disclosures

The authors declare no conflict of interest.

References

1. Rosenthal, R., Mcgranahan, N., Herrero, J. & Swanton, C. Deciphering Genetic
   Intratumor Heterogeneity and Its Impact on Cancer Evolution. *Annu. Rev. Cancer*
   Biol. 2017 **1**, 223–240 (2016).

2. Marusyk, A., Almendro, V. & Polyak, K. Intra-tumour heterogeneity: A looking glass
   for cancer? *Nature Reviews Cancer* vol. 12 323–334 (2012).

3. Landau, D. A., Carter, S. L., Getz, G. & Wu, C. J. Clonal evolution in hematological
   malignancies and therapeutic implications. *Leukemia* vol. 28 34–43 (2014).

4. Anderson, K. *et al.* Genetic variegation of clonal architecture and propagating cells in
   leukaemia. *Nature* **469**, 356–361 (2011).

5. Notta, F. *et al.* Evolution of human BCR-ABL1 lymphoblastic leukaemia-initiating cells.
   Nature **469**, 362–7 (2011).

6. Turati, V. A. *et al.* Chemotherapy induces canalization of cell state in childhood B-cell
   precursor acute lymphoblastic leukemia. *Nat. Cancer* doi:10.1038/s43018-021-00219-
   3.

7. Ebinger, S. *et al.* Characterization of Rare, Dormant, and Therapy-Resistant Cells in

Acute Lymphoblastic Leukemia. *Cancer Cell* **30**, 849–862 (2016).

8.   Lutz, C. *et al.* Quiescent leukaemic cells account for minimal residual disease in

childhood lymphoblastic leukaemia. *Leukemia* (2013) doi:10.1038/leu.2012.306.

9.   Waanders, E. *et al.* Mutational Landscape and Patterns of Clonal Evolution in

Relapsed Pediatric Acute Lymphoblastic Leukemia. *Blood Cancer Discov.* **1**, 96–111

(2020).

10.   Dobson, S. M. *et al.* Relapse-Fated Latent Diagnosis Subclones in Acute B Lineage

Leukemia Are Drug Tolerant and Possess Distinct Metabolic Programs. *Cancer*

Discov. (2020) doi:10.1158/2159-8290.cd-19-1059.

11.   Kirschner-Schwabe, R. *et al.* Chemotherapy and mismatch repair deficiency

cooperate to fuel TP53 mutagenesis and ALL relapse. *Nat. Cancer*

doi:10.1038/s43018-021-00230-8.

12.   Buitenkamp, T. D. *et al.* Acute lymphoblastic leukemia in children with Down

syndrome: A retrospective analysis from the Ponte di Legno study group. *Blood* **123**,

70–77 (2014).

13.   Xavier, A. C. & Taub, J. W. Acute leukemia in children with down syndrome.

Haematologica vol. 95 1043–1045 (2010).

14.   Schuh, A. *et al.* Monitoring chronic lymphocytic leukemia progression by whole

genome sequencing reveals heterogeneous clonal evolution patterns. *Blood* **120**,

4191–4196 (2012).

15.   Stiehl, T. & Marciniak-Czochra, A. Computational Reconstruction of Clonal

Hierarchies From Bulk Sequencing Data of Acute Myeloid Leukemia Samples. *Front.*

Physiol. **12**, 1–13 (2021).

16.   Dobin, A. *et al.* STAR: Ultrafast universal RNA-seq aligner. *Bioinformatics* **29**, 15–21

(2013).

17.   Liao, Y., Smyth, G. K. & Shi, W. FeatureCounts: An efficient general purpose program

for assigning sequence reads to genomic features. *Bioinformatics* **30**, 923–930

(2014).

18.   Wang, L., Wang, S. & Li, W. RSeQC: Quality control of RNA-seq experiments.

Bioinformatics **28**, 2184–2185 (2012).

19.   Ewels, P., Magnusson, M., Lundin, S. & Käller, M. MultiQC: Summarize analysis

results for multiple tools and samples in a single report. *Bioinformatics* **32**, 3047–3048

(2016).

20.   Love, M. I., Huber, W. & Anders, S. Moderated estimation of fold change and

dispersion for RNA-seq data with DESeq2. *Genome Biol.* **15**, 1–21 (2014).

21.   Laurenti, E. *et al.* The transcriptional architecture of early human hematopoiesis

identifies multilevel control of lymphoid commitment. *Nat. Immunol.* **14**, 756–763

(2013).

22.   Geng, H. *et al.* Self-Enforcing Feedback Activation between BCL6 and Pre-B Cell

Receptor Signaling Defines a Distinct Subtype of Acute Lymphoblastic Leukemia.

Cancer Cell **27**, 409–425 (2015).

23.   Korotkevich, G. *et al.* Fast gene set enrichment analysis. (2016) doi:10.1101/060012.

24.   Liberzon, A. *et al.* The Molecular Signatures Database Hallmark Gene Set Collection

In Brief Through extensive automated and manual curation, Liberzon et al. provide a

refined and concise collection of '"hallmark"' gene sets from the Molecular Signatures

Database for gene set enrichment analysis. Cell Systems The Molecular Signatures

Database Hallmark Gene Set Collection. *Cell Syst.* **1**, 417–425 (2015).

25.   Tirosh, I. *et al.* Single-cell RNA-seq supports a developmental hierarchy in human

oligodendroglioma. *Nat. Publ. Gr.* **539**, (2016).

26.   Hong, D. *et al.* Initiating and cancer-propogating cells in TEL-AML1-associated

childhood leukemia. *Science (80-.).* **319**, 336–339 (2008).

27.   Zur Wiesch, P. S., Engelstädter, J. & Bonhoeffer, S. Compensation of fitness costs

and reversibility of antibiotic resistance mutations. *Antimicrob. Agents Chemother.* **54**,

2085–2095 (2010).

28.   Vazquez-Santillan, K., Melendez-Zajgla, & J., Jimenez-Hernandez, & L., Martínez-
Ruiz, G. & Maldonado, & V. NF-κB signaling in cancer stem cells: a promising
therapeutic target? doi:10.1007/s13402-015-0236-6.

29.   Sergushichev, A. A. An algorithm for fast preranked gene set enrichment analysis
using cumulative statistic calculation. *bioRxiv* 060012 (2016).

30.   Dick, J. E. Stem cell concepts renew cancer research. *Blood* **112**, 4793–4807 (2008).

31.   Kreso, A. & Dick, J. E. Evolution of the cancer stem cell model. *Cell Stem Cell* **14**,
275–291 (2014).

32.   Ailles, L. E. & Weissman, I. L. Cancer stem cells in solid tumors. *Curr. Opin.*
Biotechnol. **18**, 460–466 (2007).

33.   Rehe, K. *et al.* Acute B lymphoblastic leukaemia-propagating cells are present at high
frequency in diverse lymphoblast populations. *EMBO Mol. Med.* **5**, 38–51 (2013).

34.   Shlush, L. I. *et al.* Tracing the origins of relapse in acute myeloid leukaemia to stem
cells. *Nature* **547**, 104–108 (2017).

35.   Velten, L. *et al.* Identification of leukemic and pre-leukemic stem cells by clonal
tracking from single-cell transcriptomics. *Nat. Commun.* **12**, 1–13 (2021).

36.   Lapidot, T. *et al.* A cell initiating human acute myeloid leukaemia after transplantation
into SCID mice. *MacKinnon, R. Yallen. G. Sci.* **362**, 31–39 (1993).

37.   Dick, J. E. & Bonnet, D. Human Acute Myeloid Leukaemia is organised as a heirarchy
that originates from a primitive haematopoetic cell. *Nat. Med.* **3**, 730–737 (1997).

38.   Taussig, D. C. *et al.* Leukemia-initiating cells from some acute myeloid leukemia
patients with mutated nucleophosmin reside in the CD34- fraction. *Blood* **115**, 1976–
1984 (2010).

39.   Ford, A. M. *et al.* Fetal origins of the TEL-AML1 fusion gene in identical twins with
leukemia. *Med. Sci.* **95**, 4584–4588 (1998).

40.   Greaves, M. F., Maia, A. T., Wiemels, J. L. & Ford, A. M. Leukemia in twins: Lessons
in natural history. *Blood* **102**, 2321–2333 (2003).

41.   Gale, K. B. *et al.* Backtracking leukemia to birth: Identification of clonotypic gene

fusion sequences in neonatal blood spots. *Proc. Natl. Acad. Sci. U. S. A.* **94**, 13950–

13954 (1997).

42.   Ford, A. M. *et al.* Protracted dormancy of pre-leukemic stem cells. *Leukemia* **29**,

2202–2207 (2015).

43.   Böiers, C. *et al.* A Human IPS Model Implicates Embryonic B-Myeloid Fate Restriction

as Developmental Susceptibility to B Acute Lymphoblastic Leukemia-Associated

ETV6-RUNX1. *Dev. Cell* **44**, 362-377.e7 (2018).

44.   Wray, J. P. et al. Cell cycle corruption in a pre-leukemic ETV6-RUNX1 model exposes

RUNX1 addiction as a therapeutic target in acute lymphoblastic leukemia. *bioRxiv*

Prepr. (2021).

Figure legends

Figure 1: Study of clonal dynamics through serial transplantation assays. A.

Experimental setup. BM samples before and after serial transplantation in NSG-mice were

analyzed by SNP-arrays. **B.** SNP array results of patient 1. Deletions on Chromosomes 1,

12 and a single amplification on Chromosome 17 were identified and shown to be stable

during serial transplantation. Relationships between transplants recipients are indicated with

arrows. The colour scale ranges from red (amplification) through white (neutral, diploid) to

blue (deletion). **C.** SNP array results of patient 2. Deletions on Chromosomes 9, 14 and a

single amplification on Chromosome X were identified. During serial transplantation copy

number abnormalities (CNA) disappeared and re-appeared suggesting the co-existence of

different sub-clones. Relationships between transplants recipients are indicated with arrows.

The colour scale ranges from red (amplification) through white (neutral, diploid) to blue

(deletion). **D.** Initial reconstruction of the clonal hierarchy of patient 2 leukaemia considering

Chromosomes 9, 14 and X. * Indicate secondary transplants from a primary transplant for

which SNP-array data are unavailable due to insufficient numbers of cells for analysis.

Figure 2: Detailed analysis on chromosome 9 reveals higher clonal complexity (Patient 2). A. In depth analysis of CNA on chromosome 9 during serial transplantation suggests a more complex clonal hierarchy with multiple deletions of different size targeting the locus and occurrence of cnLOH. Each box indicates an individual mouse. **B.** Phylogenetic reconstruction of leukaemia cells based on SNP-array data obtained at diagnosis and during serial transplantation. Mosaicim of CNA suggested the existence of sub-clones within analyzed samples. The panel represents a summary of the minimal phylogenetic trees that can explain the mutational landscape identified in the engrafted leukaemias. Minimal clone numbers were obtained using a computational algorithm (see supplementary materials). There exist 30 possible minimal trees, each consisting of 16 clones. The red nodes correspond to clones that are found in all 30 minimal trees. The green and orange nodes correspond to clones that are part of 24 and 6 out of 30 minimal trees respectively. The edges indicate the phylogenetic relation among the clones. The width indicates in how many of the 30 minimal trees the respective edge is present. A: loss of chromosome 9, B: short homozygous deletion on chromosome 9, B2: intermediate homozygous deletion on chromosome 9, B3: long homozygous deletion on chromosome 9, C: CnLOH on chromosome 9, D, F: copy number losses on chromosome 14, E: trisomy X, trisomy 21 (present in all subclones).The presented graph depicts the minimum number of diagnostic clones required to explain the mutational landscape identified in the engrafted leukaemias. Minimal clone numbers were calculated by computational modelling (see supplementary materials).

Figure 3: Correlation of phenotype and genotype. A. Experimental setup. BM bulk, stemB (CD34$^+$CD38$^{-/low}$CD19$^+$) and proB (CD34$^+$CD38$^+$CD19$^+$) cells were analyzed by SNP-arrays before and (if available) after serial transplantation in NSG mice. **B.** In patient 2 bulk, stemB and proB cells showed identical CNA. Deletions are shown in red and amplifications

in blue. **C.** In Patient 3 animals transplanted with bulk, stemB, and proB cells showed

identical CNA profiles: suggesting the existence of a single dominant clone. * Indicate

secondary transplants from a primary transplant for which SNP-array data are unavailable

due to insufficient numbers of cells for analysis. Secondary transplants from StemB and bulk

cells did not engraft. Deletions are shown in red and amplifications in blue.

Figure 4: Phenotypically and genetically distinct sub-clones co-exist in a case of BCR-

ABL1$^+$ ALL (Patient 4). A. FISH analysis of leukemic subpopulations at diagnosis and after

serial xenotransplantation using a BCR/ABL-specific probe. To confirm the leukemic status,

sorted cells were analysed by FISH for the appropriate clonal marker. Presence of the

t(9;22) translocation is indicated by detection of one green, one red and two red-green fusion

signals. **B.** Competitive transplants of genetically and phenotypically distinct BCR-ABL$^+$ proB

(CD34$^+$CD38$^+$CD19$^+$) and BCR-ABL$^-$ stemB (CD34$^+$CD38$^{-/low}$CD19$^+$) cells. Input and output

ratios of the two populations in individual mice are displayed. **C.** Cell cycle analysis of

leukemic StemB and proB cells from competitively transplanted NSG-mice with

subpopulations from a BCR-ABL1$^+$ ALL patient. Percentages of cells in G$_0$ (DAPI$^-$Ki67$^-$), G$_1$

(DAPI$^-$Ki67$^+$) and S/G$_2$/M (DAPI$^+$Ki67$^+$) are shown.

Figure 5: Selective therapy resistance of CD34$^+$CD38$^{-/low}$CD19$^+$ StemB cells in high risk

BCR-ABL1$^+$ ALL (Patient 4). A. Flow cytometry analysis of BCR-ABL1$^+$ ALL subpopulations

at diagnosis, remission, and relapse (patient 4). Percentages of CD34$^+$CD38$^{-/low}$CD19$^-$

(HSC), CD34$^+$CD38$^{-/low}$CD19$^+$ (stemB), CD34$^+$CD38$^+$CD19$^+$ (proB) cells and

CD34$^+$CD38$^+$CD19$^-$ (Progenitors) in total bone marrow (BM) mononuclear cells (MNC) are

shown. The black boxes show the gates used for frequency calculations. Not all collected

events are shown. **B.** Flow cytometry analysis of BM cells from serially transplanted NSG

mice injected with sorted stemB and proB cells from patient 4 relapse. **C.** PCA analysis

based on bulk RNA sequencing of diagnostic and relapse cells harvested from primary and

secondary xenografts. Six different populations of cells were analysed. i) diagnostic stemB

cells, ii) diagnostic proB cells iii) relapse stemB cells harvested from secondary xenografts

injected with stemB cells isolated from primary xenografts iv) relapse proB cells harvested

from secondary xenografts injected with stemB cells isolated from primary xenografts, v)

relapse proB cells harvested from secondary xenografts injected with proB cells isolated

from primary xenografts, and vi) relapse stemB cells harvested from secondary xenografts

injected with proB cells isolated from primary xenografts. Arrows showing the contribution of

different gene expression signatures (calculated as per panel D) to the sample's clusters

biology are displayed. **D.** Gene set enrichment analysis (GSEA) for a selection of MSigDB-

50 Hallmark gene-sets and published signatures associated with predefined differentiation

stages of the B-cell ontogeny[21]. ProB diagnostic cells are compared to the stemB diagnostic

cells. Significance thresholds: <0.05 after adjusting for multiple testing (using the Benjamini-

Hochberg method).

Figure 6: Distinct sub-clones in the BCR-ABL1$^+$ ALL and the BCR-ABL$^-$ relapse clones

are genetically related (Patient 4). A. SNP array results of diagnostic stemB, xenografted

stemB, diagnostic proB, xenografted proB, diagnostic preB, relapse stemB, relapse proB

and MRD-negative Remission sample are displayed. A chromosome 12 deletion was

present in all but the remission control sample illustrating the clonal relationship of all

leukemic samples. CNA of diagnostic stemB and proB were identified and stable during

serial transplantation with identification of some mosaicism suggesting sub-clones in each

compartment. Relationships of transplants are indicated. Relapse stemB and proB were

genetically identical but stemB showed 90% mosaicism of all clones suggesting the

existence of a pre-leukemic clone. Deletions are shown in red and amplifications in blue.

Figure 7: Pre-leukemic and leukemic hierarchy (Patient 4). Based on genetic mosaicism

and different sizes of chromosome 7 and 14 deletions a pre-leukemic clonal hierarchy could

be constructed which showed clonal evolution at the pre-leukemic that gave rise to the

diagnostic leukemic sub-clones as well as evolving in another pre-leukemic sub-clone

responsible for relapse. Based on secondary transplantation assays, relapse but not

diagnostic cells displayed phenotypic plasticity.

Supplementary Figures Legends

Figure S1: Flow cytometry sorting strategy. Representative gating strategy used for the

analysis of immunophenotypically-distinct Down's syndrome ALL subpopulations before and

after xenotransplantation. StemB (CD34$^+$CD38$^{-/low}$CD19$^+$) and proB (CD34$^+$CD38$^+$CD19$^+$)

cells were sorted. **A-B**. Patient 2 and 3 leukaemia respectively.

Figure S2: Cell cycle analysis. A. Cell cycle analysis of representative xenografted stemB

and proB as compared to the **B.** patient relapse.

Figure S3: RNA sequencing analysis. A-C Gene Set Variation Analysis (GSVA) analysis

of RNA sequencing data looking at the expression of cell cycle related gene signatures. Six

different populations of cells were analysed: i) diagnostic stemB cells, ii) diagnostic proB

cells iii) relapse stemB cells harvested from secondary xenografts injected with stemB cells

isolated from primary xenografts iv) relapse proB cells harvested from secondary xenografts

injected with stemB cells isolated from primary xenografts, v) relapse proB cells harvested

from secondary xenografts injected with proB cells isolated from primary xenografts, and vi)

relapse stemB cells harvested from secondary xenografts injected with proB cells isolated

from primary xenografts. Within each heatmap samples are sorted by category and genes

split into top 500 most variable vs all others. **A.** G1/S gene signature[25], **B.** G2/M gene

signature[25], **C.** E2F targets gene signature[24] **C.** Gene set enrichment analysis (GSEA) for a

selection of MSigDB-50 Hallmark gene-sets[24] and published signatures associated with

predefined differentiation stages of the B-cell ontogeny[21]. This analysis looks specifically at

the top 500 genes driving clustering along the PC2 component of the PCA analysis. **D.** PCA

analysis based on bulk RNA sequencing of diagnostic and relapse cells harvested from

primary and secondary xenografts. Arrows showing the contribution of gene expression

signatures involved in BCR signalling and B-cell differentiation[22] are shown. Significance

thresholds: <0.05 after adjusting for multiple testing (using the Benjamini-Hochberg method).

Figure 1

Figure 2

A

B

Connection present in
- ▬ 100%
- ▬ 80%
- ▬ 60%
- — 40%
- — 20%

of all graphs

Node present in
- **100%** (red)
- **80%** (green)
- **20%** (orange)

of all graphs

Figure 3

Figure 4

Figure 5

Figure 6

Figure 7