NK cells shape the clonal evolution of B cell leukaemia by IFN-γ production

Michelle C. Buri[1], Mohamed R. Shoeb[1], Aleksandr Bykov[1], Peter Repiscak[1], Hayeon Baik[1], Alma

Dupanovic[1], Faith O. David[1], Boris Kovacic[1], Faith Hall-Glenn[1], Jos Urbanus[2], Lisa Sippl[1], Susanne

Stofner[1], Dominik Emminger[1,3], Jason Cosgrove[4], Manfred Lehner[1,3], Leïla Perié[4], Ton N. Schumacher[2,5],

Dagmar Gotthardt[6], Florian Halbritter[1], Eva M. Putz[1,7,#]

[1] St. Anna Children's Cancer Research Institute (CCRI), Vienna, Austria

[2] Division of Molecular Oncology & Immunology, Oncode Institute, Netherlands Cancer Institute, Amsterdam, The

Netherlands

[3] Christian Doppler Laboratory for Next Generation CAR T Cells, Vienna, Austria

[4] Institut Curie, Université PSL, Sorbonne Université, CNRS UMR168, Laboratoire Physico Chimie Curie, Paris, France

[5] Department of Hematology, Leiden University Medical Center, Leiden, The Netherlands.

[6] Institute of Pharmacology, University of Veterinary Medicine Vienna, Vienna, Austria

[7] Center for Physiology and Pharmacology, Institute of Pharmacology, Medical University of Vienna, Vienna, Austria

Correspondence to Eva M. Putz, Institute of Pharmacology, Medical University of Vienna, Waehringerstrasse 13A, 1090

Vienna, Austria, eva-maria.putz@meduniwien.ac.at

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Abstract

The term cancer immunoediting describes the dual role by which the immune system can suppress and promote tumour growth and is divided into three phases: elimination, equilibrium and escape. The role of NK cells has mainly been attributed to the elimination phase. Here we show that NK cells play a role in all three phases of cancer immunoediting. Extended co-culturing of DNA barcoded mouse BCR/ABL^{p185+} B acute lymphoblastic leukaemia cells with NK cells allowed for a quantitative measure of NK cell-mediated immunoediting. Whereas most tumour cell clones were efficiently eliminated by NK cells, a certain fraction of tumour cells harboured an intrinsic resistance. Furthermore, DNA barcoding revealed tumour cell clones with secondary resistance, which stochastically acquired resistance to NK cells. We found that the production of interferon-γ (IFN-γ) rather than direct cytotoxicity by NK cells, led to the emergence of highly resistant tumour cells. Besides well-known regulators of immune evasion, our analysis of NK resistant tumour cells revealed the upregulation of novel genes, such as *Ly6a*, which drives NK cell resistance in leukaemic cells. Our results demonstrate that tumour cells are actively edited by NK cells during the equilibrium phase and use different avenues to escape NK cell-mediated eradication.

Main/Introduction

Cancer immunoediting refers to the dynamic interaction between immune cells and cancer cells during tumour development[1]. The immune system can have both tumour-promoting and tumour-suppressive effects. This process occurs in three phases: elimination, equilibrium, and escape. In the elimination phase, adaptive and innate immune cells, particularly cytotoxic T and NK cells, recognise and eliminate cancer cells. The equilibrium phase involves the coexistence of tumour cells and the immune system, leading to sculpting of the tumour under constant immune pressure, ultimately resulting in immune escape[2].

NK cells are innate lymphoid cells with the ability of recognising and eliminating damaged, stressed, and infected cells including cancer cells[3]. NK cell activation relies on a balance of activating and inhibitory signals from receptors on their surface during the interaction with potential target cells. On the one hand, activated NK cells can kill target cells through exocytosis of cytotoxic granules containing perforin and granzymes[4], induction of death receptor-mediated apoptosis[5] or antibody-dependent cellular cytotoxicity (ADCC) via their FcγRIIIA (CD16) receptor[6]. On the other hand, NK cells secrete cytokines such as interferon-γ (IFN-γ), which activate other immune cells in their vicinity[7,8]. In the context of cancer immunoediting, NK cells play a crucial role in eliminating tumour cells during the elimination phase and limiting the metastatic spread of tumour cells. NK cells spare healthy tissue and present a safe treatment option with minimal risk of causing adverse effect such as cytokine release syndrome or graft versus host disease[9]. As a result, there is growing interest in using NK cells for immunotherapies, which will benefit from a better understanding of how tumour cells evade NK cell attacks to effectively target their Achilles heel. However, most studies to date have focused on the interaction between human NK cells and tumour cell lines obtained from patients[10–15]. These tumour cells have already successfully evaded from the patient´s immune system and may thus not represent suitable models to study immunoediting.

Here, we used a model based on non-edited NK cell-naïve mouse BCR/ABL^{p185+} B cell acute lymphoblastic leukaemia (B-ALL) cells to investigate the primary interaction of tumour cells with NK

cells. This model involves a long-term co-culture of tumour cells and primary NK cells. By analysing the

NK cell-resistant tumour cells that emerge during the co-culture, we investigated molecular changes

using RNA sequencing (RNA-seq) and assay for transposase-accessible chromatin sequencing (ATAC-

seq). The results indicate that the transcriptomic changes in the resistant tumour cells are mainly due

to IFN-γ secretion by NK cells, which upregulates genes like lymphocyte antigen 6 complex, locus A

(*Ly6a*) and phospholipase A and acyltransferase 3 (*Plaat3*). This study identified Ly6A as novel potential

driver of NK cell-mediated immune evasion of leukaemia. Moreover, we established a model to

quantify NK cell effector functions and their ability to edit DNA barcoded tumour cells. Cellular DNA

barcoding stably introduces unique inheritable DNA sequences that allows tracking of individual cell

clones and their fates within a heterogeneous cell population[16–18]. Thereby, we observed that, while

some tumour cells were inherently resistant or sensitive to NK cells, a few tumour cells were edited by

NK cells to become resistant. In summary, this study shows that NK cells play an important role in all

three phases of cancer immunoediting and contribute to the active sculpting of tumour cell clones,

ultimately driving tumour evasion.

Results

Quantitative measure of NK effector functions

As a model to study NK cell evasion of leukaemic tumour cells we established a long-term *in vitro* co-

culture system using IL-2-activated primary mouse NK cells co-incubated with NK-naïve DNA barcoded

B-ALL cell lines for up to 20 days (**Figure 1A**). We generated several independent B-ALL cell lines from

the bone marrow of wild-type (n=10) and major histocompatibility complex 1 (MHC-1)-deficient $B2m^{-}$

$^{/-}$ mice (n=10) and characterised their NK receptor ligand surface expression and their susceptibility

towards NK cell-mediated eradication at an effector-to-target (E:T) ratio of 10:1. Summarizing this data

by unsupervised hierarchical clustering (**Figure 1B**) highlighted the two wild-type cell lines #13 and #15

which showed the highest susceptibility towards NK cells and expressed enhanced levels of activating

NKG2D-ligand Rae1 and lower levels of inhibitory ligands (e.g. classical and non-classical MHC-1

proteins H2-K[b], H2-D[b], Qa-1[b] and Qa-2). *B2m*-deficient B-ALL cell lines showed a higher susceptibility

towards NK cells according to the recognition of missing self[19] (see **Figure S1A** for full characterisation).

To allow for a systematic quantification of NK cell effector functions, a DNA barcode library was

introduced into the wild-type B-ALL cell lines #13 and #15 (n=2 each) and the *B2m*[-/-] (n=1) cell line #23

to track individual tumour cell clones over time. For this purpose, a new lentiviral barcode library

(LG2.1, containing 52,645 barcodes in the reference list) was generated consisting of 21 random

nucleotides positioned downstream of a green fluorescent protein (*GFP*) gene (**Figures S1B-E**). The

lentiviral library was introduced into the B-ALL cell lines at a low multiplicity of infection (MOI) to

ensure the integration of a single barcode per cell as previously described[20,21] (**Figure 1C**). To enrich for

barcoded cells, GFP[+] cells were FACS sorted shortly after transduction and before the long-term co-

culture with IL-2-activated NK cells. The barcode diversity of the resulting four cell lines called A and B

(derived from #13), C and D (derived from #15) was determined by site-specific PCR amplification and

next generation sequencing (**Figure 1D**). The barcoded cells were co-incubated with IL-2-activated NK

cells at an E:T ratio of 1:1. After 4 and 14 days, GFP[+] tumour cells were FACS purified and fresh NK cells

were replenished to re-challenge the tumour cells. Throughout the experiment, the absolute number

of GFP[+] tumour cells was determined by regular flow cytometric analysis (**Figure 1E**). While the tumour

cells initially showed significantly slower proliferation in the presence of NK cells during the first two

rounds of co-culture, this difference became marginal in the third round. Twenty-nine days after the

start of co-culture, we performed a classical 4-hour cytotoxicity assay with tumour cells that had been

co-cultured with NK cells for 20 days versus NK-naïve tumour cells. As shown in **Figure 1F**, extended

co-culture of NK and leukaemic cells induced resistance towards NK cell-mediated killing and led to

tumour evasion *in vitro*.

NK cell-mediated editing of leukaemic cells

There are four conceivable fates of B-ALL tumour cell clones upon long-term co-culture with NK cells:

The abundance of one clone could be lower ("eliminated"), higher ("primary resistant"), unchanged

("static") or highly variable ("secondary resistant") (**Figure 2A**). After bulk DNA sequencing of the

barcodes, we designed a decision tree to discriminate these four categories algorithmically (**Figure 2B**).

Primary resistance, also known as intrinsic resistance[22], is defined as *a priori* resistance towards

therapy. In our study, primary resistance was characterised by a reproducible and significant

accumulation despite the presence of NK cells in co-culture. In comparison, secondary resistance

describes the phenomenon of acquired resistance, represented by a situation in which a tumour can

initially respond effectively to therapy but relapse or progress after treatment[23]. We hypothesised that

acquired secondary resistance was not a tumour-intrinsic property but acquired only in a subfraction

of sibling cells and would be observed in isolated wells (**Figure 2A**). Indeed, clustering the DNA

barcoded leukaemic cells according to our four pre-defined categories (**Figure 2C, S2A** for summary

data), showed that the majority of leukaemic cell clones was efficiently eliminated (58.5±1.4%

considering n=2-3 independent experiments comprising n=2-4 B-ALL cell lines, mean±SEM). Only a

smaller number of leukaemic cell clones seemed to have an NK cell resistant phenotype (14.3±1.3%).

About 23.9±1.4% of the B-ALL cell clones remained static despite the presence of NK cells. As

hypothesised, we also observed B-ALL cell clones that showed secondary resistance, characterised by

the presence of the cell clone in only 1/3 replicate wells (3.3±0.8%, **Figure 2D**). To be able to draw

conclusions about the pre-existing fate of the tumour cell clones, we compared two independent

experiments with the same barcoded cell lines. As shown in **Figures 2E and S2B**, the cell clones

attributed to the clusters primary resistant and eliminated were found to have the same fate in both

experiments. In contrast, the pool of secondary resistant tumour cell clones was distinctive for

independent experiments. In summary, our data suggest that, whereas primary resistant and

eliminated cells harbour tumour-intrinsic properties *a priori* that determine their destiny, the fate of

secondary resistant tumour cell clones is stochastic. This underscores our hypothesis that secondary

resistant tumour cells acquire their resistant phenotype during NK co-culture. Our data clearly show

that apart from direct killing, NK cells participate in immunoediting by actively shaping the

tumourigenicity of individual tumour cells.

Molecular changes observed in NK cell resistant B-ALL cells

To identify novel drivers of NK cell resistance we investigated the changes in the transcriptomic and chromatin accessibility profiles of the NK cell-resistant B-ALL cell lines generated during the long term *in vitro* co-culture. To this end, we performed RNA (**Figures 3A-D, S3A-E**) and ATAC sequencing (**Figures 3E-F, S4A-C**) and compared NK co-cultured to NK naïve tumour cells. The transcriptomic analysis showed a significant upregulation of genes with a well-known link to NK cell immune evasion such as several components of the MHC-I machinery (e.g., *Stat1*, *B2m*, *Tap1* and *H2-K1*[24]) (**Figure 3A, Figures S3C-F**). Additionally, this analysis revealed a novel set of genes, which may include potential drivers of NK cell resistance, such as *Ly6a* and *Plaat3*. The functional enrichment analysis showed that the main biological process associated with the differentially expressed genes was the response to IFN-γ (e.g., *Stat1*, *Irf1*, *Ccl5* and *Cd74*) (**Figure 3B**), which is produced by activated NK cells in co-culture. After integrating the results from the four barcoded cell lines A-D, we observed that many differentially expressed genes upon NK cell co-culture were cell line specific (up: A/B=193 and C/D=99, down: A/B: 211 and C/D=63). However, a total of 129 genes were up- and 27 were downregulated in all four cell lines and may be universally associated with or driving NK cell escape (**Figure 3C**). The top-30 differentially regulated genes are depicted in the heat map in **Figure 3D**. In line, the ATAC sequencing results showed the opening of chromatin at the promoters of classical interferon-inducible genes (e.g., *Iigp1*, *Ifi27*, *Mx1*), as well as *Ly6a* and *Plaat3*, genes previously observed in the RNA sequencing dataset (**Figure 3E, S4C**). We considered overlapping differentially accessible regions (DARs) in cell lines A/B and C/D as the most robust ones (**Figure 3F**). Integration of RNA and ATAC sequencing data by comparing the differentially expressed genes and accessible regions in cell lines A/B showed a significant intersection (37 genes up/open, 16 genes down/closed). Interestingly, the two highly upregulated genes *Ly6a* and *Plaat3* in the transcriptomic analysis were also present in this strictly filtered data set (**Figure 3G**). The quantification of differential transcription factor (TF) activity by integrating the RNA- and ATAC-seq data sets revealed key TFs involved in NK cell resistance. We classified TFs as activators (NF2L1, SMRC1, JUN, IRF1 and IRF9) or repressors (RXRB, JUND), based on

the expression of the corresponding TF and accessibility of predicted binding sites[25]. STAT1 expression

was clearly upregulated but appears to have activating and repressive functions and is therefore

classified as undetermined (**Figure 3H**). A closer look at the two top hits *Ly6a* and *Plaat3* showed that

both have multiple predicted binding sites for STAT1, IRF1 and IRF9 based on the occurrence of DNA

sequence motifs in the vicinity of accessible chromatin (**Figure 3I**). In summary, multiomics analyses

revealed that the B-ALL cells, which were not *a priori* sensitive and eliminated by NK cells, showed a

highly NK cell-resistant phenotype characterised by a strong IFN-γ signature.

IFN-γ as major driver of cancer immunoediting

The two main functions of NK cells are the direct lysis of tumour cells and the production of pro-

apoptotic and -inflammatory cytokines, most importantly IFN-γ. Thus, we aimed to discriminate the

relative contributions of killing and cytokine production to cancer immunoediting. To determine the

share of direct cytotoxicity, the long-term co-culture experiment was performed with NK cells

proficient (WT) or deficient for the pore-forming protein perforin (*Prf1*$^{-/-}$). This deficiency renders them

unable to directly kill their target cells[26]. In parallel, B-ALL cells were treated continuously with

recombinant IFN-γ. As described above, the growth of GFP$^+$ tumour cells was measured by flow

cytometry during the experiment. On day 4 and 14, GFP$^+$ tumour cells were sorted and co-cultured

with freshly prepared WT NK or *Prf1*$^{-/-}$ NK cells or treated continuously with IFN-γ. As previously

observed, B-ALL cells co-cultured with WT NK cells showed an initial susceptibility during the first 14

days and developed an NK resistance thereafter (**Figure 4A**). As expected, *B2m*$^{-/-}$ B-ALL tumour cells

did not survive the co-culture with WT NK cells (**Figure S6A**). In contrast, B-ALL and *B2m*$^{-/-}$ B-ALL cells

co-cultured with *Prf1*$^{-/-}$ NK cells or treated with IFN-γ showed a minor but consistent growth inhibition

compared to untreated B-ALL cells. This showed that NK cell-derived cytokines, and in particular IFN-

γ, had a direct effect on the growth of B-ALL cells (**Figures 4A and S6A**). However, the cytotoxicity assay

performed on day 28 after treatment start showed that B-ALL cells continuously treated with IFN-γ

were most resistant to WT NK cells (**Figures 4B and S6B**).

On day 14, the B-ALL cells were further analysed for their clonal diversity and transcriptomic profiles.

In line with our previous experiments, the barcode diversity of the B-ALL + WT NK cell condition was

markedly decreased (**Figure 4C**). The tumour cells co-cultured with $Prf1^{-/-}$ NK cells or treated with IFN-

γ showed a milder decrease of tumour cell diversity (**Figure 4C**). Interestingly, we observed a strong

overlap of eliminated and primary resistant tumour cell clones in all three treatment groups, WT NK,

$Prf1^{-/-}$ NK cell and IFN-γ, suggesting that these cell clones were intrinsically susceptible or resistant to

IFN-γ-mediated apoptosis (**Figure 4D**). Apart from the primary resistant tumour cells, we detected cell

clones, which acquired resistance due to the NK co-culture or IFN-γ treatment (resistant (2nd)).

Strikingly, the number of secondary resistant tumour cell clones was the largest in the B-ALL + WT NK

cell condition (n=56 versus 15 and 12 cell clones in B-ALL + $Prf1^{-/-}$ NK or IFN-γ, respectively). These

findings led us to conclude that NK cells exert an immunological pressure on tumour cells, which is

partly independent of IFN-γ production (**Figures 4D&E**). To define the molecular changes driving NK

cell resistance in B-ALL cells upon co-culture with WT or $Prf1^{-/-}$ NK cells or treatment with IFN-γ, we

compared the transcriptomic profiles of all conditions on day 14 after start of treatment. As expected,

IFN-γ treatment induced a large number of genes, which overlapped with some of the most significant

hits of WT NK and $Prf1^{-/-}$ NK cells (23 up- and 10 downregulated genes), such as *Ly6a* and *Plaat3*. We

reasoned that changes in gene expression unique for the WT NK cell condition were due to the direct

NK cell killing. For instance, we identified an upregulation of *Bnip3*, which was recognised as important

player in immune evasion from NK cells before[27]. Additionally, *Aldoa* and *Pgk1* were found to be

upregulated, both of which are involved in glycolysis and associated with worse prognosis in

pancreatic, cervical and breast cancer[28,29] (**Figures 4F&G and S5**). In conclusion, IFN-γ on the one hand

contributed to the selective pressure of NK cells, but on the other hand prolonged IFN-γ exposure led

to the dominance of tumour cells that were less sensitive to NK cell attack.

Role of PLAAT3 and Ly6A in tumour evasion

NK cell resistance in B-ALL cells was associated with altered expression of well-known and novel genes

involved in cancer immunoediting. We demonstrated in multiple experiments that *Plaat3* and *Ly6a*

belong to the most significantly upregulated genes in NK resistant B-ALL cell lines. Therefore, we

decided to follow up on these two genes and unravel their role in immune evasion. We performed the

co-culture experiment as described in Figure 4 and analysed the B-ALL cells on day 14 by mass

spectrometry to confirm the upregulation of PLAAT3 and Ly6A on the protein level. Indeed, PLAAT3

and Ly6A protein expression in B-ALL cells was increased after WT or *Prf1*[-/-] NK cell co-culture and even

more upon IFN-γ treatment (**Figures S7A&5A**). We further generated *Plaat3* KO B-ALL clones by

CRISPR/Cas9 genome editing and verified the lack of *Plaat3* mRNA by qPCR (**Figure S7B**). *Plaat3* KO

cells did not show any difference in proliferation and viability, IFN-γ production by NK cells and NK

receptor ligand expression (**Figures S7C-F**). Moreover, loss of *Plaat3* did not have any impact on direct

NK cell-mediated killing of mouse B-ALL cells (**Figure S7G**). To test the relevance of the human

orthologue *PLAAT3* in leukaemia, we analysed the *PLAAT3* expression of acute lymphoblastic

leukaemia (ALL) patients and observed a trend of worse overall survival in patients with increased

PLAAT3 expression (**Figure S7H**). These results suggest that while PLAAT3 expression is upregulated by

IFN-γ and correlates with NK cell resistance, it does not constitute a driver of NK cell resistance.

Likewise, we deleted *Ly6a* in B-ALL cells and verified the lack of Ly6A surface protein expression by

flow cytometry (**Figure 5B**). Ly6A deficiency had no impact on B-ALL cell proliferation or survival

(**Figures 5C&D**). Although IFN-γ production by NK cells as well as surface ligand expression relevant for

NK cell interaction was unchanged (**Figures 5E&F**), *Ly6a* KO clones were more susceptible to NK cells

compared to WT clones in cytotoxicity assays (**Figure 5G**). This result clearly shows that Ly6A plays an

important role in NK immune evasion of mouse B-ALL cells. There are several close orthologues of Ly6A

in the human system. Besides a newly described human LY6A/Ly6S gene[30,31], Ly6E and Ly6K have been

well-described as close relatives in the human system. We compared the overall survival of ALL

patients with a high versus low expression of *LY6E* and *LY6K*, respectively (**Figure 5H&I**). Indeed, a high

LY6E or *LY6K* expression correlated significantly with poor survival in ALL patients.

In summary, our functional analysis of PLAAT3 and Ly6A showed that their expression is upregulated

by IFN-γ. Although PLAAT3 was persistently upregulated in our multiomics analysis, it proved not to

have any effect on NK cell resistance of the tumour cells. In contrast, Ly6A was identified as a novel

driver of tumour evasion from NK cells, by increasing the susceptibility of *Ly6a* deficient tumour cells

to NK cells. The poor disease outcome in ALL patients with high *LY6E* and *LY6K* expression supported

these findings.

Discussion

NK cells represent a potent arm of the immune system, tasked with eliminating anomalous cells.

Nonetheless, they can fail controlling cancer cells which successfully evaded from the immune

surveillance. The presented study aimed to elucidate the evasion mechanisms employed by tumour

cells to escape NK cells during initial encounter and extends earlier efforts using patient derived

tumour cell lines[10,11,15,32–36]. Notably, patient-derived cell lines have already undergone the cancer

immunoediting process and thus have acquired strategies to evade the immune system. On the

contrary, our study is based on heterogeneous, non-edited and immune-naïve mouse BCR/ABL[p185+] B-

ALL tumour cell lines. This unique model enables to explore the primary interplay between tumour

cells and the immune system. Furthermore, the B-ALL tumour cells have been DNA barcoded, a

technique that allows for the quantification of tumour cell clonal dynamics, which can help to better

understand the mechanisms of NK cell effector function. To the best of our knowledge, this particular

aspect has not been demonstrated in previous studies.

We established an *in vitro* model, whereby mouse NK and DNA barcoded B-ALL cells were co-cultured

for several days. Transcriptome analysis, assessment of chromatin accessibility and evaluation of clonal

evolution were performed at two time points during the experiment. DNA barcoding allowed the

determination of tumour cell fate during the co-culture and revealed that most tumour cell clones

possessed inherent sensitivity or resistance to NK cells. This technique has already been used by many

others to quantify tumour cell susceptibility towards immune cells, drug- or antibody-based

immunotherapies[37–40]. Similar to our research, these studies observed that the majority of tumour cells

is predestined for being sensitive or resistant to treatment. For instance, Baldwin et al. found pre-

existing resistant clones which were positively selected after combined anti-PD1 and anti-CTLA4

therapy in mouse mammary cancer and their gene expression signature correlated with worse survival

in patients[39]. Our study highlighted tumour cell clones beyond those two fundamental fates of life and

death, which acquired secondary resistance in the presence of NK cells. Secondary resistance was only

observed in a small subset of tumour cell clones, which showed NK cell resistance in only one of three

replicate wells on day 14 of co-culture. Similarly, using DNA barcoding to investigate the clonal

evolution of drug-treated lung cancer cells, Acar *et al.* recently described the emergence of "*de novo*

resistant lineages" in one replicate only[37]. Interestingly, we found that secondary resistant tumour cells

were highly distinctive for each independently performed experiment, suggesting that a stochastic

event causes the resistance. This demonstrates that NK cells can shape tumour cells and contribute to

immunoediting of tumour cell clones. However, due to the necessity of expanding the barcoded cell

lines before the experiments (to ensure reproducible representation of all barcodes in every well), we

cannot formally exclude the possibility of pre-existing heterogeneity within a cell clone. Furthermore,

the barcode sequence of the LG2.1 library was found to be located approximately 900 bps upstream

of the polyA tail and could thus not be detected using single-cell RNA sequencing. It would certainly be

interesting to delve deeper into secondary resistance mechanisms and to compare the transcriptomic

profiles of primary and secondary resistant tumour cell clones in the future[41,42].

We show that NK cells exert immunological pressure on tumour cells, which eventually escape and

become refractory to NK cell killing. Amongst others, previously described NK cell tumour evasion

mechanisms include the upregulation of proteins involved in antigen presentation and MHC-I and -II

molecules[43,44], the loss or shedding of NKG2D ligands such as MHC class I chain-related molecule A and

B (MICA/B)[45] and secretion of TGF-β and IL-10[46]. Notably, we here describe *Ly6a* overexpression as

emerging novel driver of tumour evasion. Ly6A, also called stem cell antigen-1 (Sca-1), is an 18-kDa

mouse glycosyl phosphatidylinositol-anchored cell surface protein. It plays an important role in

haematopoietic progenitor/stem cell lineage fate, brain blood barrier transport and cancer stem cell

biology[47–50]. Ly6A was shown to be upregulated in cancer stem cells and many tumour entities and is

induced by Wnt/β-Catenin signalling and TGF-β deregulation. This promotes cell migration *in vitro,*

increased tumourigenicity and resistance to chemotherapy *in vivo*[50–53]. The human LY6 gene family

comprises several members, but the direct orthologue of mouse *Ly6a* has been under debate for a

long time. Very recently, two independent groups described a novel protein-encoding gene located in

region 8q24.3 of the human genome called *LY6A/LY6S,* which is upregulated upon IFN-γ treatment and

shows high expression in pituitary tumours[30,31]. Previously, other *LY6* family members were suggested

as close homologues of the murine Ly6a gene, such as *LY6D*, *LY6E*, *Ly6H* and *LY6K*, which were

implicated as novel biomarkers for poor cancer prognosis and are frequently amplified in human

cancer[54]. Al Hossiny *et al.* showed that human LY6E and LY6K promote breast cancer *in vivo* and drive

drug sensitivity and epithelial-to-mesenchymal transition. Moreover, they observed that LY6E/LY6K

are required for TGF-β signalling and facilitate immune evasion by upregulating PD-L1 on cancer cells,

by recruiting T-regulatory cells and dampening NK cell activation[55]. In line, our survival analysis of a

paediatric ALL-patient cohort confirmed the cancer driving feature of LY6E and LY6K also in leukaemia.

Additionally, our results support the observed oncogenic and immune suppressive properties of Ly6A

showing that it is highly upregulated after NK cell co-culture and inducible by IFN-γ. Although the exact

mechanism remains unknown, the sensitivity towards NK cells increased significantly upon *Ly6a* gene

deletion. Our findings align with the notion that immunoedited and thus more aggressive tumour cells

may harbour cancer stem cell (CSC) features. CSCs are considered responsible for metastasis and

therapy resistance, as they are characterised by deregulated differentiation, self-renewal, drug

resistance, and amongst others evasion of immunosurveillance. Unlike other immune cells, NK cells

are believed to show direct cytotoxicity against CSCs[56]. On the other hand, as CSCs frequently

upregulate classical and non-classical MHC-I molecules while losing NK cell activating ligands, it is still

a matter of debate if and how NK cells are able to detect and kill CSCs[56,57].

In general, this research underscores the pivotal role of IFN-γ in the acquisition of NK cell resistance

and as ultimate driver of immune evasion. In contrast, *in vitro* and *in vivo* gene-editing screens showed

that loss of IFN-γ signalling (IFNGR1/2, JAK1/2, etc.) and antigen presentation (B2M, TAP1, etc.) leads

to immune evasion from T cells or resistance to immunotherapies[58–60]. However, downregulation of

MHC-I converts the tumour cells to prime NK cell targets. It was thus proposed to treat tumours with

an NK cell-based therapy to antagonize immune escape from T cells[34]. In our experiments, tumour cells

responded to the NK cell co-culture with an upregulation of many IFN-γ inducible genes, such as MHC-

I molecules, thereby inhibiting NK cells and inducing resistance. We sought to uncover the role of IFN-

γ by comparing the tumour cells co-cultured with WT NK cells to tumour cells treated with IFN-γ. By

using DNA barcoding we were able to quantify the effect of IFN-γ on tumour cells and dissect IFN-γ

dependent and independent tumour cell fates. We observed that a major part of NK cell susceptible

clones overlapped well with the eliminated clones upon IFN-γ treatment (572 of 795 clones). We thus

concluded that IFN-γ plays on the one hand a major role in tumour cell elimination and on the other

hand enhances NK cell resistance of the remaining tumour cells by upregulating IFN-γ dependent

genes, such as genes involved in MHC-I representation and *Ly6a*. The pro-apoptotic functions of IFN-γ

have been observed previously in diverse tumour types. IFN-γ signalling was recognised to induce

apoptosis-related genes such as caspase 3 and caspase 7 mainly via the JAK/STAT1 pathway[61–64].

Although IFN-γ has profound antiproliferative and proapoptotic effects against cancer cells, IFN-γ

treatment of cancer patients achieved only minor success probably due to its poor therapeutic index[65].

Additionally, it has been proposed that IFN-γ has a dual role in cancer[66]. Its anti- and protumorigenic

activities are dependent on the duration (acute versus chronic), magnitude of IFN-γ signalling and the

immune cell composition. A deeper understanding of the multifaceted actions of IFN-γ is needed to

provide patients the most promising treatment options.

In conclusion, our work sheds light on the intricate interplay between NK and leukaemic cells during

initial encounters. This study shows that leukaemic cells are actively shaped by NK cells and unravels

novel cancer evasion strategies by identifying potential driver genes. Further research, especially *in*

vivo, is required to delve deeper into the tumour evasion mechanisms involving Ly6A. These findings

ultimately deepen our understanding of the initial interaction of tumour and NK cells and eventual

cancer immune evasion.

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Figures

Figure 1

Figure 1: Co-culture system to study cancer immunoediting *in vitro*. (A) Scheme illustrates the

experimental setup: BCR/ABL[p185+] B-ALL leukaemic cells (LCs) are transduced with a DNA barcode

library. Barcoded LCs are co-cultured with NK cells for 4-20 days *in vitro*. **(B)** Ten B-ALL cell lines were

generated from C57BL/6 WT mice and characterised regarding the surface expression of NK ligands

(black) and their susceptibility towards NK cell killing (red, depicted here is the cytotoxicity at E:T=10:1)

measured by flow cytometry. Cell lines #13 and #15, in blue, showed the highest NK cell susceptibility

and were used for the following experiments. Rows were centered; unit variance scaling was applied

to rows. Both rows and columns were clustered using correlation distance and average linkage. The

plot was generated using the Clustvis web tool[67]. **(C)** B-ALL cell lines were DNA barcoded by lentiviral

transduction. Using different amounts of viral supernatant, the optimum transduction efficiency

(<5%)[16] was chosen to allow a single barcode integration per cell as determined by GFP expression and

measured by flow cytometry. Shown are means±SD (n=2) of one representative cell line. **(D)** Cell lines

A-D were analysed for their barcode diversity. Two B-ALL cell lines (#13 and #15) were barcoded (n=2

independent transductions, <5% transduction efficiency) and the barcode diversity of the FACS-

purified GFP$^+$ cell lines was determined by targeted DNA sequencing. Cell lines A and B were derived

from #13 and cell lines C and D from #15. Bars show barcode diversity of barcoded tumour cell lines

(A&D n=3 and B&C n=2 independent experiments, each determined by sequencing of 3 biological and

2 technical replicates). Shown is median with interquartile range. **(E)** Barcoded WT B-ALL cell lines A-D

(n=4) were cultivated in three individual wells either in the absence or presence of purified and IL-2-

activated NK cells. One *B2m$^{-/-}$* B-ALL cell line was included as control. Absolute numbers of B-ALL cells

were determined by flow cytometry. On day 0, 4 and 14, B-ALL cells were FACS sorted, re-plated and

fresh NK cells were replenished. Shown are means±SD of n=4 B-ALL cell lines of one representative

experiment (n=3 independent experiments); the significance was calculated by an unpaired t-test. **(F)**

On day 29, a standard 4h NK cytotoxicity assay was performed using B-ALL cells that had been cultured

for 20 days in the presence or absence of NK cells. Shown are means±SEM of n=4 cell lines (in technical

triplicates) of one representative experiment (n=3 experiments); the significance was calculated by a

paired t-test.

Figure 2

Figure 2: Quantification of NK cell-mediated cancer immunoediting *in vitro*. (A-B) Schematic

representation of the proposed model. Upon long-term co-culture of B-ALL and NK cells, we

hypothesised that each tumour cell clone would fall in one of the following categories: The abundance

of the clone can be significantly higher (primary resistant) or lower (eliminated), unchanged (static) or

show a high variability (secondary resistant). **(B)** A rule-based decision tree was designed to

discriminate the four categories depicted in (A). Variability is defined as $v = log2(max(x +$

$1)/min(x + 1))$, where x is a vector of normalised counts across samples. **(C)** The heatmap shows the

normalised abundance of barcodes (norm. BC abundance) of cell line A (n=3 wells) from one

representative experiment (in total: n=4 cell lines, n=2 experiments, n=3 wells, sequenced in technical

duplicates). Each column in the figure represents a sample (well), and each row represents a barcode.

The barcodes were divided into the four pre-defined groups according to the criteria defined in (B).

The Shannon index (shown above the heatmap) serves as measure of barcode diversity and drops

significantly after 14 days of NK co-culture. The histogram on the left shows the average abundance of

the barcoded cell clone in the B-ALL only samples on day 4. **(D)** The bubble plot depicts the normalised

abundance of barcodes of resistant (2nd) clones in B-ALL samples (day 4) compared to B-ALL + NK

samples (day 14). The size and colour of the bubbles indicate the normalised barcode abundance.

Shown here is the same cell line and experiment as in (C). **(E)** Comparing two individual experiments

using the same representative cell line A as shown in (C&D), the Euler diagrams highlight a high overlap

of eliminated, resistant (1st), and static cell clones. In contrast, only two resistant (2nd) clones were

shared between both independent experiments.

Figure 3

Figure 3: Integrative analysis of RNA-seq and ATAC-seq shows well-known and novel genes

deregulated in NK resistant B-ALL cells. The transcriptome **(A-D and G)** and chromatin accessibility **(E-**

I) analysis of B-ALL cells co-cultured with NK cells for 4 (time point 1 (TP1)) or 14 (TP2) days were

analysed and compared to the non-co-cultured B-ALL cells on TP1 (n=3-4 independent experiments).

(A) The volcano plot shows the significantly up- or downregulated genes in B-ALL cell lines A and B

after co-culture with NK cells for 14 days (n=3 experiments). Statistics were calculated with the Wald

test (adjusted p-value (padj)_cutoff < 0.05; log_2FC_cutoff > 0.58 (FC = 2)). **(B)** The dot plot illustrates

the enrichment analysis using the hyperR tool[68] for the Gene Ontology Biological Process (GOBP) terms

associated with NK co-cultured B-ALL cells shown in panel (A). X-axis and dot colour depicts the false

discovery rate (FDR) and y-axis the corresponding biological processes. Dot size illustrates the number

of genes associated with the biological processes. **(C)** The Euler diagrams display up- or downregulated

genes in cell lines A/B and/or C/D after 14 days of NK co-culture. **(D)** Heatmap shows the expression

of selected overlapping genes (top-20 up-regulated, top-10 down-regulated; ranked by p-value)

identified in (C) (n=3-4 experiments, 4 B-ALL cell lines). The row-scaled normalized counts represent

the log_2(fold change). **(E)** The volcano plot shows differentially accessible chromatin regions in B-ALL

cells A and B after co-culture with NK cells for 14 days (n=3 experiments). Statistics were calculated

with the Wald test (padj_cutoff < 0.05; log_2FC_cutoff > 0.58 (FC = 2)). **(F)** The Euler diagrams highlight

genes whose promoters (TSS +/- 3kb) had higher or lower chromatin accessibility in cell lines A/B

and/or C/D after 14 days of NK co-culture in n=3-4 experiments. Gene promoters with the highest

|log_2(fold change)| are highlighted. **(G)** Integration of RNA- and ATAC-seq data shows that 37 genes

are up- and 16 downregulated in cell lines A/B; the top 10 up- or downregulated genes are highlighted

on the right. **(H)** Differential transcription factors (diffTF) analysis[69] highlighted activator (green) and

repressor (red) transcription factors (TFs) that were differentially expressed in B-ALL cell lines A and B

after NK co-culture (right side of the plot) or B-ALL alone (left side of the plot). **(I)** Integrative Genomics

Viewer (IGV) genome browser tracks display representative ATAC-seq signal densities at the *Ly6a* and

Plaat3 loci in B-ALL cells with or without NK co-culture. Shown are the gene bodies including a 3 kb

upstream promoter region, as well as binding sites of TFs identified in (H). Database HOCOMOCO

(v10)[70] was used and depicted are counts per million (CPM) with a range of [0-1.37].

Figure 4: NK cell-derived IFN-γ is a major driver of leukaemia immunoediting. (A) Barcoded WT B-ALL

cell lines (n=2 cell lines A&D) were cultivated in three individual wells either in the absence or presence

of purified and IL-2-activated WT or *Prf1*⁻/⁻ NK cells or with 2 ng/ml IFN-γ. Absolute numbers of B-ALL

cells were determined by flow cytometry. On day 0, 4 and 14, B-ALL cells were FACS sorted, re-plated

and fresh NK cells were replenished. Shown are means±SD of n=2 B-ALL cell lines of one representative

experiment (n=2 independent experiments). The significance was calculated with the Kruskal-Wallis

test and Dunn´s multiple comparisons testing. **(B)** On day 28, a standard 4h NK cytotoxicity assay with

WT NK cells was performed using B-ALL cells that had been cultured for 27 days in the presence or

absence of NK cells or IFN-γ. Shown are means±SEM of 2 cell lines (in technical triplicates) of one

representative experiment (n=2 experiments). Statistics were calculated with a matched two-way

ANOVA with Dunnett´s multiple comparisons test. **(C)** B-ALL cells were FACS sorted on day 14, and the

barcode diversity was determined by targeted DNA amplicon sequencing. The Shannon diversity of

leukaemic clones dropped significantly (p = 0.0134) upon co-culture with WT NK cells, and to a lesser

degree with $Prf1^{-/-}$ NK cells or recombinant IFN-γ. Shown is the mean and dots depict each replicate.

Statistics were calculated with the Kruskal-Wallis and Dunn´s multiple comparisons testing. **(D)** The

barcodes within each sample were classified according to the decision tree in Fig. 2B. The barcodes

that were attributed to the four pre-defined groups were compared between the treatment groups.

Heatmaps indicate the number of barcodes shared between treatment groups (bright yellow = high,

dark green = low). Resistant (1st) and eliminated leukaemic clones were broadly overlapping between

WT or $Prf1^{-/-}$ NK- or IFN-γ-treated B-ALL cells. However, resistant (2nd) leukaemic clones were highly

specific for each of these conditions. **(E)** The bubble plots show the normalised abundance of barcodes

of resistant (2nd) clones in B-ALL samples compared to B-ALL + WT or $Prf1^{-/-}$ NK or IFN-γ samples (day

14). The size and colour of the bubbles indicate the barcode abundance. **(F)** The Euler diagram displays

the number up-regulated genes in cell lines A/D after 14 days of co-culture with WT or $Prf1^{-/-}$ NK cells

or stimulation with recombinant IFN-γ. The orange list highlights the top 10 upregulated genes in the

intersection of the three comparisons and the red list shows the WT NK cell specific top 10 upregulated

genes. **(G)** The Euler diagram displays the number of downregulated genes in cell lines A/D after 14

days of co-culture with WT or $Prf1^{-/-}$ NK cells or stimulation with recombinant IFN-γ. The blue list

highlights the top 10 downregulated genes in the intersection of the three comparisons and the red

list shows the WT NK cell specific top 10 downregulated genes. (A-G) Shown is one representative

experiment (A&B, n=2 cell lines) or one representative cell line from one experiment (C-G, n=2 cell

lines).

Figure 5: The role of Ly6A in the evasion of leukaemic cells from NK cell-mediated surveillance. (A)

Bar graphs show protein abundances of Ly6A on day 14 of the same conditions as in the co-culture

experiment described in Figure 4. Two B-ALL cell lines (A in circles/D in triangles) were analysed by

mass spectrometry in triplicates. Bars represent means of arbitrary units (AU) relative to the whole

protein abundance; the significance was calculated by one-way ANOVA with Tukey post-hoc test. **(B)**

Ly6a KO clones were generated by ribonucleoprotein transfection into parental B-ALL cell lines. The

gene modification was verified by TIDE (Tracking of Indels by Decomposition) PCR and sequencing (not

shown) as well as by measuring Ly6A surface expression by flow cytometry. **(C)** Growth curve of *Ly6a*

WT and KO B-ALL clones (n=2 per genotype) in absolute cell numbers over 13 days. **(D)** Cell viability of

Ly6a WT and KO clones (n=2 per genotype) over 13 days. **(E)** Measurement of IFN-γ secretion by mNK

cells upon culturing in tumour cell-conditioned medium for 48 h. As positive control mNK cells were

activated with PMA/Ionomycin (n=2 clones per genotype and all conditions were in technical

duplicates). **(F)** *Ly6a* WT and KO B-ALL cell lines (n=4 per genotype) were characterised by their surface

expression of NK ligands by flow cytometry. Columns represent different tumour cell clones ordered

according to genotype and cell line. Data has been both scaled and clustered by row to highlight

variations in marker expression across different samples. Red indicates higher and blue lower

expression levels of surface ligands. **(G)** A 4h standard NK cytotoxicity assay was performed using WT

and *Ly6a* KO B-ALL clones. The graph shows the summary data of three independent experiments.

Shown are means±SEM of n=2 *Ly6a* WT and KO clones (p-value = 1.15e-8). The p-value was calculated

utilising a t-test after fitting the specific lysis values of the 3 biological replicates on a linear model

(more details in the methods section). **(H&I)** Kaplan-Meier plots depicting survival probabilities of the

TARGET-ALL-P2 patient cohort divided into *(H)* *LY6E* and **(I)** *LY6K* high and low expressing groups (cut-

off percentile = 60% and 20%, respectively). Graph was generated by using the online web tool:

cSurvival (ubc.ca).

Methods

Mice

C57BL/6JRj wild-type mice were obtained from the Janvier labs and C57BL/6-Prf1[tm1Sdz]/J (*Prf1*[-/-]) mice[26] were obtained from the Jackson Laboratory (Strain #002407). B6.129-B2m[tm1Jae] (*B2m*[-/-])[71] mice were kindly provided by Thomas Kolbe (VetMedUni Vienna). The mice were between 8 and 12 weeks of age and maintained under specific pathogen-free conditions according to FELASA guidelines.

Cell culture

To generate BCR/ABL[p185+] (B-ALL) cell lines, bone marrow was isolated from 8-10-week-old mice and single cell suspensions were prepared through repeated homogenisation using 18G and 24G needles. 48 hours before the infection of bone marrow cells, Phoenix-ECO cells (ATCC) were transiently transfected with a retroviral MSCV-BCR/ABL[p185] vector (kindly provided by Veronika Sexl) mixed with PureFection transfection reagent (System Biosciences). The bone marrow cells were transduced for 1 hour with the fresh viral supernatant in the presence of 10 μg/ml polybrene (Sigma Aldrich) and 10 ng/ml recombinant murine IL-7 (Peprotech). Transformed B-ALL cell lines were maintained in RPMI medium (Thermo Fisher Scientific) supplemented with 100 U/ml penicillin and 100 μg/ml streptomycin (Thermo Fisher Scientific), 10% fetal bovine serum (FBS) (Capricorn) and 50 μM β-mercaptoethanol (Sigma-Aldrich) (cRPMI).

Primary mouse NK cells were isolated from spleens of 8-12-week-old wild-type or *Prf1*[-/-] mice using the negative selection EasySep mouse NK cell isolation kit (STEMCELL Technologies) or positive selection CD49b (DX5) MicroBeads kit (Miltenyi Biotec) and cultured in cRPMI supplemented with 5,000 U/ml recombinant human IL-2 (Miltenyi Biotec). Mouse NK (mNK) cell purity was measured by flow cytometry with the fixable viability dye eFluor 780 (Invitrogen) and the following antibodies purchased from Miltenyi Biotec: mCD335 (REA815), mCD45.2 (REA1223), mNK1.1 (REA1162) and mCD3 (REA641). mNK cells were identified by gating on single viable mCD45.2[+] mCD3[-] NK1.1[+] mCD335[+] cells.

Phoenix-ECO and Lenti-X packaging cells (Takara) were cultured in DMEM medium (Thermo Fisher Scientific) with 10% FBS.

697 All cell lines were regularly tested for their mycoplasma negativity by using and following the

698 manufacturer´s instructions of the PCR Mycoplasma detection kit (ABM). Cell growth and survival was

699 measured by regular cell counting using a CASY counter (OLS OMNI Life Science).

DNA barcoding

Generation of the lentiviral barcode library LG2.1

The BC1DS_lib oligo containing a 21nt random barcode sequence (**Table S1**) was PCR amplified (14

cycles: 10sec 98°C, 30 sec 57°C, 20 sec 72°C) with Phusion polymerase (New England Biolabs (NEB)).

The resulting PCR amplified product was column purified (MinElute PCR cleanup kit, Qiagen) and

subsequently digested with XhoI and AscI, followed by ligation into the 3' UTR of the GFP cDNA

sequence within the pRRL lentiviral vector, using the Electroligase kit (NEB). Electrocompetent

ElectroMax Stbl4 bacteria (Invitrogen) were then electroporated with 6 ng ligation product and a small

fraction of the transformed bacteria was plated on Luria-Bertani (LB) agar plates to determine

transformation efficiency, while the remaining bacteria were grown overnight in 400 ml LB medium

(VWR Life Science) supplemented with 100 µg/ml ampicillin (Sigma Aldrich). DNA was extracted from

the bacterial culture using the Maxiprep kit (Invitrogen).[17,72] This barcode library was kindly provided

by Ton Schumacher and amplified further to generate our in house library. ElectroMax Stbl4 bacteria

were transformed with 10 ng barcode library to achieve a transformation rate of 100 times the

estimated barcode diversity of the library. Bacteria were cultured on LB agar medium (bacteriological

agar, tryptone, yeast extract (Oxoid) and sodium chloride (Roth)) petri dishes with 100 µg/ml ampicillin

and on the next day resuspended in 50 ml LB medium (tryptone, yeast extract and sodium chloride)

with 100 µg/ml ampicillin. Bacteria were expanded in up to 400 ml liquid culture. DNA was extracted

with EndoFree Plasmid Maxi Kit (Qiagen). To determine the library's barcode diversity and distribution,

barcodes were amplified with a three PCR reaction protocol (see below) in biological and technical

duplicates and sequenced on a HiSeq 3000/4000 device (Illumina) with 64 bp single-end reads at the

Biomedical Sequencing Facility (BSF) Vienna.

Barcoding of tumour cells

Lenti-X cells were transfected with the LG2.1 barcode library and the two packaging plasmids p8.9QV

(provided by Ton Schumacher) and VSVG (provided by Veronika Sexl) using the PureFection

transfection reagent. Virus supernatant was harvested after 48 hours and 72 hours and concentrated

with AMICON columns (Sigma Aldrich). 2×10^5 B-ALL cells were spinfected with serially diluted viral

supernatant for 90 minutes at 900xg. The transduction rate (% GFP$^+$ cells) was measured by flow

cytometry 3 days after transduction. Tumour cells with the desired transduction rate (<5%[16]) were

expanded and FACS sorted to enrich GFP$^+$ cells. Barcoded tumour cells were frozen for further

experiments.

Flow cytometry

***In vitro* co-culture system and tumour cell growth analysis**

9×10^5 barcoded B-ALL cells were co-cultured with mNK cells in a effector-to-target (E:T) ratio of 1:1 in

three biological replicates per condition. Depending on NK cell cytotoxicity and tumour cell growth,

cells were harvested for FACS sorting and further analyses (described below) on day 4 or 6 and for the

second time point on day 14 or 16. On the day of harvesting, 3×10^5 tumour cells were co-cultured for

a second or third time with fresh mNK cells (E:T = 1:1). All conditions were cultured in the presence of

2,500 U/ml rhIL-2. Wherever indicated, tumour cells treated with 2 ng/ml mouse IFN-γ (Abcam) every

2nd to 3rd day, while splitting or expanding the tumour cells.

During the co-culture, tumour cell growth was monitored by flow cytometry on a FACSymphony A5

Cell Analyzer device (Becton Dickinson). The following antibodies (clones) were purchased from

Miltenyi Biotec: mCD45.2 (REA1223), mNK1.1 (REA1162), mCD3 (REA641) and mCD19 (REA749).

Zombie Red (Biolegend) was used as viability dye. The absolute number of B-ALL cells was determined

with AccuCheck Counting Beads (Invitrogen). Tumour cells were identified by gating on single live

mCD45.2$^+$ mCD19$^+$ GFP$^+$ cells. NK cells were detected by gating on single live mCD45.2$^+$ NK1.1$^+$ mCD3$^-$

cells.

Surface marker analysis of tumour cells

Tumour cells were split to 3×10^5 cells/ml two days before the experiment and 10^6 cells were used for

the staining. The following antibodies (clones) were used: CD48 (REA1238), CD112 (829038), CD137L

(REA962), CD154 (REA785), CD155 (REA519), TRAIL-R (CD261) (REA1233), Qa-1b (6A8.6F10.1A6), Qa-

2a (REA523), H-2Db (REA1198), H-2Kb (REA619), Rae1 (REA723) and Ly6a/Sca-1 (REA422) purchased

from Miltenyi Biotec. Single living cells were gated according to size and granularity in FSC-A, SSC-A

and FSC-H plots and cell surface marker expression was quantified by median fluorescence intensity

(MFI).

Cytotoxicity assay and effector function assay

Cytotoxicity was measured with a flow cytometry-based assay. 5×10^4 CellTrace Violet (Invitrogen)

stained B-ALL cells were seeded in 96-well round bottom plates and IL-2-activated mNK cells added in

different E:T ratios (10:1, 5:1, 1:1) in technical duplicates or triplicates. After 4 hours of incubation at

37°C, cells were washed with PBS, stained with fixable viability dye eFluor 780 (Invitrogen) and fixed

with fixation buffer (Biolegend). Subsequently, samples were measured with a HTS plate reader on an

LSR Fortessa device (Becton Dickinson). Percentage of specific NK cell lysis was calculated as follows:

(% dead tumour cells after co-incubation with NK cells - % spontaneous lysis) / (100 % - % spontaneous

lysis).

Multi-omics approaches

10^6 GFP$^+$ barcoded tumour cells were FACS sorted, washed twice with PBS and snap frozen. DNA and

RNA were extracted using the AllPrep DNA/RNA Mini Kit (Qiagen).

DNA barcode sequencing

Library preparation for barcode sequencing was performed in technical duplicates with three nested

PCR steps. PCR#1 amplified the barcodes, PCR#2 added the required sequences for the universal

Illumina adapters and four random base pairs (bps) (BC1v2DS For) for a more diverse sequencing start.

PCR#3 attached the P5 and P7 universal Illumina adapters and with the P7 a sample-specific index

sequence (Primer sequences are in **Table S1**). The PCR#3 product fragment size (322 bps) was

confirmed on a 2% agarose gel electrophoresis for every sample. The samples were cleaned up with

Agencourt AMPure XP beads (Beckman Coulter). Up to 96 equimolar pooled samples spiked with 10-

15% PhiX (Illumina) were sequenced with the MiSeq Reagent Kit v3 (Illumina) 64 bp single-end reads

with an aimed coverage of 50 to 100 reads/barcode.

Barcode/Cell clone evolution analysis

The R statistics software v4.0.3 was used to analyse barcode abundance and clonal diversity.

Unmapped reads were loaded from FASTQ format (function *readDNAStringSet* from package

Biostrings v2.58.0[73]) into the R environment. First, we located the barcode sequences by pattern

matching (function *vmatchPattern* from package Biostrings v2.58.0) to identify the fixed head

("GAACACTCGAGATCAG") and non-variable ("TGTGGTATGATGT") portions of the reads, which directly

enclose the barcodes from left and right, respectively. We then extracted the stretch of sequence in

between. Sequences that did not meet the expected length (21 bases) or included uncalled bases ("N")

were discarded (mean discarded sequences = 15%, **Table S2**). To define the barcodes of the reference

library (**Figure S1C-E**), we selected sequences with a sum of at least 100 reads across samples (n = 4)

of the viral library for further analysis. With these we constructed a count matrix indicating the number

of reads matching each barcode. Next, the technical replicates were merged by adding up the read

counts. For each cell line, only barcodes detected across three B-ALL alone samples at the first time

point (TP1) or at the second time point (TP2) with a total sum of at least 9 reads were considered for

downstream analysis (**Table S3**). Barcode diversity for each sample was quantified using Shannon

diversity as follows $sum(p * log(p), na.rm = TRUE) * -1$, where p is the ratio of the counts of

each barcode relative to the total barcode counts. We used DESeq2[74] v1.30.0 to find the subset of

differentially abundant barcodes (abs[log_2(fold change)_cutoff] ≥ 1 & padj_cutoff < 0.05) across time

point and conditions of interest. Shifted log_2 transformation was used for normalisation (function

normTransform from package DESeq2 v1.30.). Based on the statistics of differential abundance and

variability across samples (here, the variability was calculated as following: $v = log2(max(x +$

$1)/min(x + 1))$, where x is a vector of normalised counts across samples) at every time point.

Barcodes were assigned to one of five categories: i) primary resistant: differentially abundant (padj <

0.05) and increasing (log_2(fold change) ≥ 1; comparing B-ALL + NK vs B-ALL alone); ii) eliminated:

differentially abundant (padj < 0.05) and decreasing (log$_2$(fold change) ≤ -1; comparing B-ALL + NK vs

B-ALL alone); iii) static: variability v ≤ 0.5 in B-ALL alone, and v < 1 at TP 2; iv) secondary resistant: low

variability in B-ALL alone v ≤ 0.5, high variability in B-ALL + NK TP2 v ≥ 1 and barcode abundance high

in only one of the three replicates at TP2; v) others: none of the above. Barcodes were further ordered

using hierarchical clustering (function *hclust* from package *stats* v 4.0.3, method = "ward.d2").

Heatmaps were generated using ComplexHeatmap v2.7.5.1 and bubble plots were generated using

ggplot2 v3.3.3.

RNA sequencing

After RNA isolation (see above), samples were treated with DNase I (Thermo Fisher Scientific). Library

preparation and sequencing was performed at the Biomedical Sequencing Facility (BSF) of the

Research Center for Molecular Medicine (CeMM) of the Austrian Academy of Sciences.

RNA-seq libraries were prepared with the QuantSeq 3' mRNA-Seq Library Prep Kit (FWD) for Illumina

(Lexogen). Library concentrations were quantified with the Qubit 4.0 Fluorometric Quantitation

system (Thermo Fisher Scientific) and the size distribution and RNA integrity number were assessed

using the 2100 Bioanalyzer instrument (Agilent). For sequencing, samples were diluted and pooled into

NGS libraries in equimolar amounts. Expression profiling libraries were sequenced on HiSeq 3000/4000

or Novaseq 6000 instruments (Illumina) following a 50-base-pair and 120-base-pair single-end setup,

respectively. Raw data acquisition and base calling was performed on-instrument, while the

subsequent raw data processing off the instruments involved two custom programs based on Picard

tools (2.19.2). In a first step, base calls were converted into lane-specific, multiplexed, unaligned BAM

files suitable for long-term archival (IlluminaBasecallsToMultiplexSam, 2.19.2-CeMM). In a second

step, archive BAM files were demultiplexed into sample-specific, unaligned BAM files

(IlluminaSamDemux, 2.19.2-CeMM).

RNA-seq analysis

The obtained raw reads were processed using the Snakemake pipeline[75]. Mapping of the reads was

performed using STAR[76] (v2.7.10a) to the Genome Reference Consortium Mouse Build 38 (GRCm38),

and duplicate reads were identified with Picard (v2.27.4). Quantification of the mapped reads was

carried out using Salmon[77] (v1.8.0). The following data processing and visualisation was performed in

R (v4.2.0). To ensure robustness of the analysis, the data set was filtered to include only transcripts

detected in at least three samples. The analysis of cell lines A and B (both established from cell line

#13) was performed separately from the analysis of cell line C and D (both originating from cell line

#15). Finally, the results were compared.

For count normalisation, variance stabilization, and differential expression analysis, the Bioconductor

DESeq2[74] (1.38.3) package was used. DESeq2 employs a negative binomial distribution-based model,

with additional shrinkage using apeglm[78] algorithm (v1.20.0). Transcripts were considered significantly

differentially expressed between groups of comparison if they exhibited a $\log_2$(fold change) ≥ 0.58 and

a padj ≤ 0.05. Only these transcripts were selected for subsequent analysis steps.

The initial exploratory analysis included principal component analysis (PCA), production of expression

heatmaps and volcano plots using ggplot2 (v3.4.2) and heatmap (v1.0.12). To correct for batch effects

between different experiments in PCA plots (**Figures S3A&B, S4A**) the limma R[79] package (v3.54.2) was

utilised. The resulting gene lists were annotated and subjected to gene set enrichment analysis (GSEA).

Visualisation of the GSEA results was achieved using the hypeR[68] package (v2.0.1) and represented as

dot plot. Gene Ontology Biological Process (GOBP) terms were obtained from the Molecular Signatures

Database (MSigDB) using the msigdbr[80] (v7.5.1) package.

ATAC sequencing

5×10^4 GFP$^+$ tumour cells were FACS sorted into 1.5 ml micro centrifuge tubes containing 400 µl PBS

(Sigma Aldrich) and 0.5% Bovine Serum Albumin (BSA, GE Healthcare) and pelleted. Cell lysis and DNA

tagmentation was performed in a total volume of 25 µl of 2x TD (Tagment DNA Buffer) and TDE1

(Tagment DNA enzyme) (Illumina), 1% Digitonin (Promega) and 50x EDTA-free Protease Inhibitor

Cocktail (Roche) for 30 minutes at 37°C. Samples were purified with the Qiagen MinElute kit (Qiagen).

Libraries were generated as described by Buenrostro et al[81] by using the Ad1_noMx and Ad2.1-Ad2.48

indexed primers (**Table S1**). PCR products were purified with Agencourt AMPure XP beads and sample

profile was analysed with the High Sensitivity D5000 kit (Agilent Technologies) on a TapeStation

(Agilent Technologies). 45-48 equimolar pooled samples were sequenced on a NovaSeq 6000

instrument with 50 bp paired-end reads at the BSF (CeMM).

ATAC-seq analysis

The obtained raw reads were processed using the ATAC-seq Next Flow[82] pipeline (v2.0). This included

mapping the reads with STAR to the GRCm38 genome, identification of duplicates with Picard

(v2.27.4), calling broad peaks with MACS2[83] (v2.2.7.1). Peaks overlapping with blacklisted (GRCh38)

regions were filtered out and the remaining peaks were annotated with HOMER[84] (v 4.11). Consensus

peaks across all samples were identified with BEDTools[85] (v2.30.0) and counting reads in these peaks

was done with featureCounts[86] (Rsubread v2.0.1). Visualisation of the ATAC-seq peaks was performed

by creation of BigWig files with bedGraphToBigWig (v377) and the following export to IGV[87] browser

(v2.16.1). Quality of the obtained raw data was assessed with MultiQC[88] (v1.13). The following data

processing and visualisation was performed in R (v4.2.0). To ensure robustness of the analysis, the data

set was filtered to include only consensus peaks detected in all three replicates of each condition and

cell line and only the peaks that were located in the window of ±3kb around transcription start site

(TSS). The analysis of cell lines A and B was performed separately from the analysis of cell line C and D.

Similar to the RNA-seq analysis, the Bioconductor DESeq2 (1.38.3) package was used for counts

normalisation, variance stabilisation, and analysis of differential accessibility. Peaks were considered

significantly accessible between groups of comparison if they exhibited a $\log_2$(fold change) ≥ 0.58 and

a padj ≤ 0.05. Exploratory analysis included PCA, production of heatmaps, volcano plots and GSEA.

Integration of RNA-seq and ATAC-seq data

874 To integrate two datasets and perform genome-wide quantification of differential transcription factor

875 activity in cell lines at different time points we used diffTF[69] (v1.8). The processing was done only for

876 A and B data sets, due to very high computational demand of the algorithm. The input consisted of

877 raw ATAC-seq reads in BAM files, filtered consensus peaks from the previous step, raw RNA-seq counts,

878 and PWM data for all TFs – HOCOMOCO[70] (v10).

2D-RP/RP Liquid Chromatography – Tandem Mass Spectrometry analysis

Cells were lysed in 200 µL lysis buffer (50 mM HEPES pH 8.0, 2% SDS, 1 mM PMSF, cOmplete Protease

Inhibitor Cocktail, Roche), heated to 99 °C for 5 min, and DNA was sheared using a Covaris S2

ultrasonicator. Lysates were cleared at 20,000 × g for 15 min at 20 °C and tryptic digest was performed

using FASP according to Wisniewski et al[89]. Peptides were desalted (C18 SPE; Pierce peptide desalting

columns, Thermo Fisher Scientific), labelled with TMTpro™ 16plex reagents according to

manufacturer´s instructions (Thermo Fisher Scientific), pooled, dried in a vacuum concentrator, and

cleaned up via C18 SPE. After re-buffering in 10 mM ammonium formiate buffer pH 10, peptides were

fractionated via a C18 column (150 × 2.0 mm Gemini-NX, 3 µm, C18 110Å, Phenomenex, Torrance, CA,

USA) on a Dionex UltiMate 3000 nanoLC (Thermo Fisher Scientific). Fractions were dried in a vacuum

concentrator and taken up in 0.1% TFA before analysis.

Mass spectrometry analysis was performed on an Orbitrap Fusion Lumos Tribrid mass spectrometer

(Thermo Fisher Scientific) coupled to a Dionex UltiMate 3000 nanoLC system (Thermo Fisher Scientific)

via a Nanospray Flex Ion Source (Thermo Fisher Scientific) interface using the same procedure as Block

et al[90]. Data analysis was performed using the Proteome Discoverer v.2.4.1.15 platform.

Genome editing

Ribonucleoprotein (RNP)-mediated CRISPR genome editing of B-ALL cells

RNP electroporation: crRNAs, Alt-R CRISPR Cas9 tracrRNA ATTO 550 and Alt-R S.p. Cas9-GFP V3 were

purchased by Integrated DNA Technologies (IDT). Cells were transfected according to manufacturer´s

instructions using the following electroporation settings: 350 V, 5 ms and 1 pulse with the Gene Pulser

Xcell electroporation system (BioRad). crRNA sequences: *Ly6a*: TCACGTTGACCTTAGTACCC;

TATTGAAAGTATGGAGATCC, *Plaat3*: CGTCATGTTTGTTATTGACC; ATGGCCCAGTGTCTGTACAT. The next

day tumour cells were FACS sorted for GFP and ATTO 550 double positivity. After assessing editing

efficiency by TIDE sequencing (described below) of the bulk population, cells were single cell FACS

sorted into 96-well flat bottom plates into 50% tumour cell conditioned medium and 50% fresh cRPMI.

After the outgrowth of single clones, the KO was validated by TIDE sequencing. Therefore, up to 2×10^5

cells were pelleted and lysed in 50 µl Pawel´s buffer (50 mM potassium chloride (Roth), 10 mM Tris-

HCl pH9 (Sigma Aldrich), 0.45% Nonidet-p40 (Calbiochem), 0.45% Tween 20 (Sigma), 0.8 mg/ml

proteinase K (Biozym) in deionized water) at 56°C for 15 minutes with subsequent heat inactivation at

95°C for 5 minutes. After centrifugation at 2000xg for 5 minutes, supernatant was used for the TIDE

PCR to amplify the breakpoints of the guide RNAs (primer sequences in **Table S1**). PCR products were

Sanger sequenced (Microsynth) and analysed for their genotype with the TIDE (nki.nl) web tool. Gene

knockouts were validated using flow cytometry (described above) or quantitative PCR (described

below).

Quantitative PCR

For the validation of the *Plaat3* KO in tumour cells, RNA of 2×10^6 cells was extracted with the Monarch

Total RNA Miniprep Kit with "on column" DNase I treatment (NEB). cDNA was generated by using

LunaScript RT SuperMix Kit (NEB) and qPCR was performed with the Luna universal qPCR Master Mix

(NEB) (primer sequences in **Table S1**).

IFN-γ ELISA

Tumour cells were split to 3×10^5 cells/ml on day 0 and cultivated for 48 hours in cRPMI. On day 2, the

supernatant was harvested, filtered through a 45 µm filter, and transferred onto IL-2 expanded mNK

cells in the presence of 2,500 U/ml IL-2. After 48 hours, mNK cells were harvested and the supernatant

was frozen at -80°C. As positive control served the stimulation of mNK cells with PMA/Ionomycin

(Invitrogen) 4 hours before the supernatant harvesting. The ELISA was performed according to

manufactures protocol (Abcam) and analysed using the analysis tool:

"https://www.arigobio.com/elisa-analysis".

Statistical analysis

Statistical analysis was performed using GraphPad Prism version 8.4.3 for Windows (GraphPad

Software; http://www.graphpad.com) using two-tailed unpaired t-tests, one-way, two-way ANOVA or

Kruskal-Wallis test as indicated. Where ANOVA showed a statistical difference, Tukey's multiple

comparison testing or Dunnett's post-hoc testing was applied. The α level for all tests was set to 0.05,

and p values were two tailed. The significance level is indicated as following: $*p < 0.05$; $**p < 0.01$;

$***p < 0.001$; $****p < 0.0001$).

To pool the data from independent cytotoxicity assays (n=3 experiments, Figures 5C&F), a linear model

with the categorical variables 'Experiment', 'E:T Ratio', 'Genotype', 'Cell line' and 'Replicate' as

predictors for the cytotoxicity response was fit using the R function 'lm' which performs a t-test to

determine significance of the relationship between the predictor variables and the response variable.

For visualisation purposes, the effect of experimental batch on the cytotoxicity response was corrected

for by subtracting the estimated linear model coefficients for experiments 2 and 3 from the observed

specific lysis values of the respective experiments, as the approach used in the 'removeBatchEffect'

function from the R package limma.

To generate the heatmap in Figure 1B, the data was pooled from two independent experiments by

calculating the mean surface marker expression or mean % specific lysis at E:T=10:1. The values were

standardised for each variable (row) using the following calculation: Z-score = (mean of observed

values – mean of the samples) / (standard deviation of the samples).

Heatmaps in Figure S6D&H were generated by using the R package ComplexHeatmap (version 2.10.0).

The raw median fluorescence intensity (MFI) values minus the MFI of their unstained control served

as input. Rows (surface markers) were scaled and clustered (unsupervised). Columns (cell lines) were

ordered by genotype and cell line.

Data and code availability

The data and codes generated in the current study are available at Github

(https://github.com/TumorImmunoEditingLab). All other relevant data are available from the

corresponding author.

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Author contributions

Conceptualization: E.M.P.

Methodology: M.C.B., B.K., E.M.P., with input from J.U., J.C., L.P., T.S., D.G.,

Formal analysis: M.C.B., M.R.S, A.B., P.R., E.M.P.

Investigation: M.C.B., H.B., A.D., F.O.D., B.K., F.H.-G., L.S., S.S., D.E., D.G.

Writing (original draft): M.C.B., E.M.P.

Visualization: M.C.B., M.R.S., A.B., E.M.P.

Resources: J.U., J.C., M.L., L.P., T.S., F.H., E.M.P.

Supervision: F.H., E.M.P.

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980   All authors reviewed and edited the manuscript.

Competing interest declaration

982   The authors declare no competing interests.

Additional information

984   **Corresponding author:** Eva Maria Putz, Center for Physiology and Pharmacology, Institute of

985   Pharmacology, Medical University of Vienna, Waehringerstrasse 13A, 1090 Vienna, Austria. E-mail:

986   eva-maria.putz@meduniwien.ac.at

Supplementary data figures

Supplementary Figure 1

Supplementary Figure 1: $B2m^{-/-}$ B-ALL cell line characterisation and details about the barcode library.

(A) Ten $B2m^{-/-}$ B-ALL cell lines were generated from B6.129-B2m^{tm1Jae} ($B2m^{-/-}$) mice and characterised regarding the surface expression of NK cell ligands and their susceptibility towards NK cell killing (depicted here is the cytotoxicity at E:T=10:1). Rows were centred; unit variance scaling was applied to rows. Both rows and columns were clustered using correlation distance and average linkage. The plot was generated using the Clustvis web tool[67]. **(B)** Plasmid vector map of the pRRL-CMV-*GFP*-BCv2AscI barcode library plasmid. In the inner circle the CMV promoter, the *GFP* gene, barcode nucleotides (n=21) with Toplib and Botlib primers and the ampicillin resistance are highlighted. Restriction enzyme sites are depicted on the outer circle. **(C)** The barcode library was deep sequenced in technical replicates after plasmid transformation, MaxiPrep and following a 3-step nested PCR library preparation protocol. After raw data processing and filtration for valid barcodes, the barcodes of the

technical replicates were compared and overlapped well. Common barcodes are shown in orange;

non-overlapping barcodes from replicates 1 or 2 are show in blue or green, respectively. **(D)** The graph

shows a representative sequence of one replicate of the barcode library (n=4). Positions 1-4 represent

the random nucleotide insertion after using the BC1v2DS_For primer in PCR#2. Positions 5-23 and 45-

65 are the constant sequences before and after the barcode, respectively. The 21 nucleotides from

positions 24-44 represent the barcode sequence. **(E)** The barcode reference library generated in-house

(Putz laboratory) was compared to the one used in the Périe laboratory. Both libraries, which were

sequenced independently, were overlapping to a high degree. The Putz library detected more

barcodes, which was likely due to the higher sequencing depth.

Supplementary Figure 2

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Supplementary Figure 2: Extended quantification of NK cell-mediated cancer immunoediting *in vitro*.

According to our hypothesis, the abundance of each barcoded tumour cell clone can be significantly

higher (primary resistant) or lower (eliminated), unchanged (static) or show a high variability

(secondary resistant) after extended NK cell co-culture. **(A)** The percentage of the cell clones in each

group on day 14 is depicted for cell lines A (as shown in Figure 2), B, C and D. Each dot represents the

summarised analysis of one independent experiment (plated in three biological replicates and

sequenced in technical duplicates). **(B)** Each experiment was performed at least two times using the

same barcoded parental cell line. The Euler diagrams compare two individual experiments of cell lines

B, C and D (A is shown in Figure 2). The tumour cell clones that were categorised as eliminated, primary

resistant, and static showed a relatively high overlap between the two independent experiments. In

contrast, secondary resistant clones were less likely to stem from the same original clone when

comparing independent experiments.

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Supplementary Figure 3: Analysis of differentially expressed genes in NK cell resistant tumour cells.

(A&B) The principal component analysis (PCA) plot illustrates the transcriptomic relatedness among

cell lines **(A)** A and B and **(B)** C and D and conditions (B-ALL alone, B-ALL + WT NK cells on time point 1

(TP1) and TP2) in the long-term co-culture experiments, summarising 3 or 4 independent experiments,

respectively. Data was batch corrected for experiments. **(C)** Volcano plot of the RNA-seq analysis of

cell lines C and D showing DEGs of B-ALL + WT NK co-cultured cells (TP2) versus B-ALL alone samples.

(D&E) The heat maps summarise DEGs of cell lines **(D)** A and B (n=3 experiments) and **(E)** C and D (n=4

experiments). Heat maps were clustered by condition, cell line and experiment. Z-score $\log_2$(fold

change) is shown and up- or downregulated genes are depicted in red and blue colour, respectively.

(F) After 20 days of culturing tumour cell line D with or without NK cells, changes in ligand expression

were measured by flow cytometry. Shown are MFIs gated on the tumour cells.

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Supplementary Figure 4

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Supplementary Figure 4: Analysis of differentially accessible genes in NK cell resistant tumour cells.

(A&B) PCA plots from the ATAC-seq analysis compares the similarities of the chromatin accessibility of

the cell lines and the different conditions. Cell lines A and B are shown in **(A)** and C and D in **(B)**

summarising 3 and 4 independent experiments, respectively. **(C)** The DARs in B-ALL + WT NK cells at

TP2 versus B-ALL alone at TP1 were compared and are shown in a volcano plot summarising 4

independent experiments using cell lines C and D.

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Supplementary Figure 5: Transcriptomic profiles of tumour cells co-cultured with WT NK cells or *Prf1*-

/- NK cells or treated with IFN-γ. (A) PCA analysis of the transcriptomic data from RNA-seq shows that

the B-ALL cells treated with IFN-γ possess the most unique transcriptome and form a cluster on their

own in both cell lines (A and D). Data was batch corrected for experiments. **(B-D)** Volcano plots showing

DEGs of the three treatment conditions (B-ALL + WT NK, B-ALL + *Prf1*-/- NK and B-ALL + IFN-γ) compared

to the B-ALL alone samples. Analysis includes both cell lines. Differentially up- or downregulated genes

are depicted in red and blue, respectively. **(E-G)** Heat maps depicting all up- or downregulated genes

of the above-mentioned comparisons. Heat maps were batch-corrected for cell line and clustered by

cell line and condition. Z-score $\log_2$(fold change) shows the expression of genes from low to high

expression depicted with a colour bar from blue to red.

Supplementary Figure 6

Supplementary Figure 6: NK cell-derived IFN-γ is a major driver of leukaemia immunoediting. (A) The

barcoded $B2m^{-/-}$ B-ALL cell line was cultivated in three individual wells either in the absence or presence

of purified and IL-2-activated WT or $Prf1^{-/-}$ NK cells or with 2 ng/ml IFN-γ. Absolute numbers of B-ALL

cells were determined by flow cytometry. On day 0, 4 and 14 B-ALL cells were FACS sorted, re-plated

and fresh NK cells were replenished. Shown is one representative experiment (n=2 independent

experiments). **(B)** On day 28, a standard 4h NK cytotoxicity assay was performed using $B2m^{-/-}$ B-ALL

cells that had been cultured for 27 days in the presence or absence of NK cells or IFN-γ. Shown are

means±SD of one cell line in technical triplicates of one representative experiment (n=2 experiments).

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Supplementary Figure 7: Generation of *Plaat3* KO B-ALL clones and their characterisation. (A) Bar

graph shows the protein abundance of PLAAT3 on day 14 of the same conditions as in the co-culture

experiment described in Figure 4. Two B-ALL cell lines (A in circles/D in triangles) were analysed by

mass spectrometry in triplicates. Bars represent means of arbitrary units (AU) relative to the whole

protein abundance; the significance was calculated by one-way ANOVA with Tukey post-hoc test. **(B)**

Plaat3 KO clones were generated by ribonucleoprotein transfection into parental B-ALL cell lines. The

gene modification was verified by TIDE (Tracking of Indels by Decomposition) PCR and sequencing (not

shown) as well as by *Plaat3* mRNA expression analysed by qPCR. Bars and error bars represent

means±SD; the significance was calculated using an unpaired t-test. **(C)** Growth curve of *Plaat3* WT and

KO clones (n=4 per genotype) showing the proliferation in absolute cell numbers over 13 days. **(D)** Cell

viability of *Plaat3* WT and KO clones (n=4 per genotype) measured over 13 days. **(E)** Measurement of

IFN-γ secretion by mNK cells upon culturing in tumour cell conditioned medium for 48 h. As positive

control mNK cells were activated with PMA/Ionomycin (n=4 clones per genotype and all conditions

were in technical duplicates). **(F)** *Plaat3* WT and KO B-ALL cell lines (n=4 per genotype) were

characterised by their surface expression of NK ligands by flow cytometry. Columns represent different

tumour cell clones ordered according to genotype and cell line. Data has been both scaled and

clustered by row to highlight variations in marker expression across different samples. Red indicates

higher and blue lower expression levels of surface ligands. **(G)** A 4h standard NK cytotoxicity assay was

performed using WT and *Plaat3* KO B-ALL clones. The graph shows the summary data of three

independent experiments. Shown are means±SEM of *Plaat3* WT and KO cell clones (n=4 per genotype)

(not significant). The p-value was calculated utilising a t-test after fitting the specific lysis values of the

3 biological replicates on a linear model (more details in the methods section). **(H)** Kaplan-Meier plot

depicting survival probability of the TARGET-ALL-P2 patient cohort divided into *PLAAT3* high and low

expressing groups (cut-off percentile = 40 %). The graph was generated by using the online web tool:

cSurvival (ubc.ca).

Supplementary table legends

Supplementary Table 1: Primers. Columns show primer name, sequence, usage, annealing temperature (in °C) and used polymerase.

Supplementary Table 2: Valid DNA barcode reads. The table shows the sample name, cell line, invalid and valid read counts. Invalid reads, which did not match the filter criteria were discarded. Valid read count divided by total read count indicates the percentage of the retrieved valid reads within a sample. Listed are EXP 9.6 and EXP 9.7 (Figure 2&S2, n=4 cell lines A-D) and EXP 9.8 (Figure 4&S2A, n=2 cell lines A and D). All conditions are in triplicates. Example sample name: "Experiment_Cell line_Condition_Timepoint_Well_Technical replicate".

Supplementary Table 3: Read counts per barcode of all samples. The table consists of sheets unique for each cell line and experiment. Each sheet shows in the rows the barcode sequences (barcode_seq) detected for the specific cell line and experiment. Columns depict each sample sequenced in technical duplicates. Shown are EXP 9.6 and EXP 9.7 (Figure 2&S2, n=4 cell lines A-D) and EXP 9.8 (Figure 4&S2A, n=2 cell lines A and D). All conditions are in triplicates. Example sample name: "Experiment_Cell line_Condition_Timepoint_Well".

Supplementary Table 4: Antibodies and viability dyes used for flow cytometric analyses. Indicated are target antigens, conjugates, clones, suppliers, catalogue numbers and used working dilutions.

Figure 1

Figure 2

A

well 1   well 2   well 3

+ NK cells

eliminated

resistant (1st)

resistant (2nd)

static

B

Is the barcode abundance significantly different in co-cultured compared to naïve tumour cells?

padj < 0.05 ?

yes   no

Clone behaves similar in all three wells

High variability in co-cultured tumour cells?

|variability| ≥ 1

decreasing   increasing   yes   no

eliminated

$\log_2(\text{fold change}) \leq -1$

resistant (1st)

$\log_2(\text{fold change}) \geq 1$

resistant (2nd)

static

C

Shannon diversity

eliminated (n=524)

resistant (1st) (n=128)

resistant (2nd) (n=50)

static (n=145)

norm. BC abundance scaled $\log_2(\text{counts}+1)$

Avg count day 0 $\log_{10}(\text{counts}+1)$

B-ALL day 4   B-ALL + NK day 4   B-ALL + NK day 14

D

Barcode/tumour cell clone

B-ALL   B-ALL + NK

norm. BC abundance scaled $\log_2(\text{counts}+1)$

E

eliminated   resistant (1st)

192 | 332 | 57   9 | 119 | 55

EXP 1
EXP 2

resistant (2nd)   static

48 | 2 | 53   73 | 72 | 67

Figure 3

Figure 5