Title

Enhancer lncRNA LOC730338 modulates BCR signaling and immune evasion in lymphoma by
regulating RNA homeostasis.

Running Title

LOC730338 modulates lymphoma immunogenicity

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Highlights

- The enhancer-associated long non-coding RNA ADARreg modulates RNA editing and alters the stability and localization of key immune-related transcripts, thereby enabling tumor immune evasion.
- ADARreg has prognostic and therapeutic relevance, linking resistance mechanisms and immune modulation, which supports the development of RNA-based therapeutic strategies in cancer.

Summary

Chronic antigenic stimulation is a central factor in the development of marginal zone lymphoma (MZL). While the pharmacological inhibition of the B-cell receptor (BCR) signaling by Bruton's tyrosine kinase (BTK) inhibitors is initially effective, the development of resistance remains a challenge in treating MZL and other B-cell malignancies. Enhancer activation remodeling is a key epigenetic mechanism that enables tumor adaptation during therapy. The most active regulatory regions cluster in super-enhancers and produce enhancer RNAs (eRNAs), a class of unstable noncoding transcripts that primarily serve as scaffolds for chromatin looping. However, when stabilized, these eRNAs can evolve into long noncoding RNA (lncRNAs) with distinct functions.

To investigate enhancer-associated long non-coding RNAs (elncRNAs) involved in shaping BCR pathway dependence, we conducted a CRISPR interference (CRISPRi) screen in MZL cells. We identified LOC730338, an elncRNA linked to A-to-I RNA editing, which we renamed ADARreg. ADARreg renders tumor cells refractory to BCR pathway inhibition by modulating ADAR2 nuclear translocation and altering RNA modification patterns in key regulatory isoforms through coordinated control of RNA stability and localization, as revealed through subcellular direct RNA sequencing (DRS). In addition, ADARreg induces an immune-suppressive transcriptional program, increasing the production of inhibitory cytokines and receptors that diminish NK cell-mediated cytotoxicity.

Together, these findings uncover a novel role for elncRNAs in orchestrating immune evasion and provide a potential therapeutic strategy to overcome resistance in lymphoma and other immune-related diseases.

Graphical abstract

Keywords

elncRNAs, CRISPRi-screen, Direct RNA Sequencing, differential isoform usage, epitranscriptomics, nuclear trafficking, tumor- microenvironment crosstalk

INTRODUCTION

Marginal zone Lymphoma (MZL) derives from MZ B cells under chronic inflammation and prolonged B
cell receptor (BCR) activation from persistent infections or autoimmune diseases [1]. A hallmark of MZL is
the interaction of tumor cells with the microenvironment, a complex network of various cell types,
including T cells, which communicate through soluble factors like cytokines and chemokines [2]. BCR
signaling plays a crucial role in shaping these interactions and activating downstream pathways like NF-
κB to sustain lymphoma growth [3,4]. Indeed, the pharmacological inhibition of BCR signaling, particularly
through Bruton's tyrosine kinase (BTK) inhibition with the first-in-class ibrutinib and newer agents, is an
effective therapeutic option for MZL patients as well as for other B-cell neoplasms and autoimmune
disorders [5,6]. However, resistance frequently develops, sustained by epigenetic reprogramming [7-11]. The
specific regulators driving this transcriptional plasticity remain incompletely understood, with growing
attention focusing on non-coding RNAs (ncRNAs), particularly enhancer-associated long non-coding
RNAs (elncRNAs), which play key roles in regulating gene expression [12,13]. ElncRNAs are transcribed from
enhancer regions and function as tissue-specific transcriptional regulators [14]. They derive from super-
enhancers (SEs), highly acetylated enhancer clusters that control lineage-defining gene programs [13,15].
Increasing evidence implicates elncRNAs in cancer [16], where their enhancer-derived origin places them
at key nodes of oncogenic transcriptional networks. By stabilizing protein complexes *in cis* or acting *in*
trans to influence nuclear and cytoplasmic processes, elncRNAs can promote tumor progression and
therapeutic resistance [17,18]. Their restricted expression patterns, tight linkage to specific enhancer
landscapes, and capacity to modulate key signaling pathways collectively make elncRNAs attractive
candidates for mechanistic investigation and potential therapeutic intervention [19].

Here, we aimed to identify elncRNAs modulating the transcriptional reprogramming in ibrutinib-
resistant splenic MZL (SMZL) via a screen based on the Clustered Regularly Interspaced
Short Palindromic Repeats Interference (CRISPRi) technology. Among the candidates, the lncRNA
LOC730338 (AC011294.3/ENSG00000233539) emerged due to its previously identified role in regulating
A-to-I RNA editing in B cells [20]. We show that this transcript likely functions as a negative regulator of
ADAR2 activity. Based on this, we renamed it ADARreg and demonstrated its role in modulating the BCR
pathway and its immune-suppressive capacity. Using direct RNA sequencing (DRS) of subcellular
fractions, we uncovered a complex regulation of RNA homeostasis, including RNA stability and
trafficking, orchestrated by ADARreg.

RESULTS

Acquired ibrutinib-resistance in SMZL cells is associated with changes in super-enhancer activation

To investigate the mechanism by which MZL cells become BCR-independent and stop responding to
anti-BCR pathway treatment, we utilized an *in vitro* model previously developed in our laboratory [21].
The VL51-Ibru is a derivative of the VL51, a *bona fide* SMZL cell line [22], in which a nine-month exposure
to increasing concentrations of the BTK inhibitor ibrutinib has led to secondary resistance to BTK
inhibitors and other BCR pathway-targeting drugs, such as the PI3K inhibitors idelalisib, duvelisib, and
copanlisib [21]. *In vitro* BCR stimulation revealed that, in the ibrutinib-resistant VL51-Ibru, NF-κB signaling
exhibited reduced BCR dependency compared to parental cells (Fig. S1A).

Since VL51-Ibru has no DNA mutations in BCR pathway genes that are recurrently mutated in patients
(including *BTK* and *PLCG2*), as reported in 21, we studied the epigenetic remodeling of the resistant and
parental cells using ChIP-sequencing to profile histone modifications (H3K27ac, H3K4me1, and
H3K4me3). We identified 33,264 H3K27ac peaks: 20,950 common, 2,830 unique to VL51, and 541
unique to VL51-Ibru (Fig. 1A, Table S1). Enhancers were defined by the concomitant presence of
H3K27ac and H3K4me1 and the absence of H3K4me3. Compared to its parental counterpart, VL51-Ibru
lost 1,329 enhancers and gained 144 new ones (Fig. 1A, B). Utilizing RNA-seq data from parental and
derivative cells, we examined the genes located within 40 kb of these differentially activated enhancers.
Loss of enhancers was linked to the downregulation of key genes, including FOXO1, CDKN2A, CDKN2B,
and MAP3K14, with the most notable decrease observed in NLRP4, a pattern-recognition receptor (PRR)
involved in immune activation (Fig.1C) [23]. A few genes near newly activated enhancers were significantly
upregulated in resistant cells (Fig.1C). These findings suggest that, in addition to regulating target genes,
newly activated enhancers may function through the non-coding transcripts associated with them.

Identification of elncRNAs associated with ibrutinib resistance

Super-enhancers (SEs) are densely bound by transcription factors and highly acetylated. They control
the correct temporal and tissue-specific gene expression and are the primary source of elncRNAs[14,24].
Using the ROSE algorithm [25], we identified 1,345 SEs from H3K27Ac ChIP-Seq data. (Fig.S1B). To
determine how many SEs were associated with elncRNA production, we reconstructed the
transcriptomes of VL51 and VL51-Ibru cells using total RNA-Seq profiling. We included VL51-Ide, a VL51
derivative resistant to PI3Kδ inhibitors [11], to expand our analysis beyond BTK inhibition resistance. By
analyzing ribosomal RNA-depleted libraries, we detected polyadenylated and non-polyadenylated
transcripts and identified novel and known transcripts associated with SEs (Fig. S1B). We defined
elncRNAs as any transcript detectable within an active SE in either parental or resistant cells. We
assigned a transcription start site (TSS) based on the FANTOM project CAGE dataset [26] (Fig. S1B). We
excluded transcripts annotated as protein-coding genes but included antisense or sense intronic
transcripts. We identified 427 elncRNAs, including 74 novel transcripts not listed in GENCODE (Fig.S1B).
Of these, 13% (45 elncRNAs) were lost, and 9% (30 elncRNAs) were gained in the VL51-Ibru cells
(Fig.S1B, Table S2).

Functional screening of elncRNAs linked to BCR blockade

To identify elncRNAs that mediate or oppose BCR blockade, we performed a genome-wide CRISPRi
screen in both parental and ibrutinib-resistant cells (Fig.2A). We designed a custom library of paired
guide RNAs (pgRNAs) that targeted 255 annotated and 57 novel elncRNAs (Fig.2B). The library also
included pgRNAs targeting 225 long intergenic ncRNAs (lincRNAs) differentially expressed between the
two cell lines, 68 essential genes as positive controls, and 33 non-expressed transcripts as negative
controls. Altogether, the library covered 660 transcripts (Fig.2B, Table S2). The pgRNAs were designed to
target regions within -300 to +300 base pairs around the TSSs by a dead Cas9 fused with the repressor
ZIM3 (dCas9-KRAB-ZIM3) (Fig.2A, S1C), to transcriptionally repress avoiding DNA damage [27].

To evaluate the impact of elncRNA silencing on BCR blockade, after elncRNA CRISPRi library infection
(Fig.S1D), parental and resistant cells were treated with either ibrutinib or DMSO for two weeks (Fig.2C).
We analyzed two replicates per condition at three time points (6, 10, and 14 days) compared to time 0

(Fig.2C). The effects of specific transcripts on cell survival and BCR dependency, assessed by ibrutinib
sensitivity, were measured through pgRNAs enrichment overtime, represented by β score (Fig.2D, Table
S3). Positive β values indicate that transcripts impaired proliferation or increased drug sensitivity,
negative β values indicate the opposite. The high β correlation for positive controls across replicates
validated the robustness of our approach (Fig.S1E).

The β correlation for elncRNAs targets was consistent, albeit higher in parental cells, possibly reflecting
greater heterogeneity in the resistant population (Fig.S1F). Multi-dimensional scaling (MDS) plots
showed that lincRNA silencing differentiated the two cell lines, especially after 14 days (Fig. S1G).
Similarly, elncRNA silencing discriminated them as early as day 6 (Fig.2D). Notably, some elncRNAs
affected resistant cell survival only during drug exposure (Fig.2D).

The distribution of β scores in parental and resistant cells exposed to ibrutinib showed concordance
among many transcripts modulating the drug response, especially among lincRNAs (Fig. 2E, left).
However, some elncRNAs peculiarly affected ibrutinib efficacy in resistant or parental cells (Fig. 2E,
right). Noteworthy, stable silencing of AC011294.3 (ADARreg) and RP11-96H19.1 (referred to lncPCED1b
by LNCipedia, or SLC38A4-AS1 by GENCODE V49 due to its proximity to these genes (Fig.S2A)) was
sufficient to revert BCR dependency in resistant cells (Fig. 2E, right).

ADARreg and lncPCED1b affect the sensitivity to ibrutinib

To validate the CRISPRi screen results, we transduced VL51-Ibru cells with constructs expressing
mCherry and pgRNAs targeting the ADARreg or lncPCED1b TSS (Fig.S2A, B) and GFP-tagged dCas9ZIM3.
As a negative control, we used pgRNAs targeting the TSS of KLK12, a gene not expressed in this model.
After two weeks, cells expressing both fluorescent markers were sorted and analyzed. Knockdown of
ADARreg or lncPCED1b increased sensitivity to BTK inhibition in ibrutinib-resistant cells (Fig. S2C). In
contrast, parental cells with lncPCED1b knockdown did not show this phenotype (Fig. S2C). Notably, in
ibrutinib-resistant cells, the phenotype remained stable over time and was maintained in monoclonal
lines generated weeks after selection of the polyclonal population (Fig. 2F). These findings highlight the
essential role of ADARreg and lncPCED1b in sustaining BCR blockade in ibrutinib-nonresponsive cells and
underscore the value of these stable models for dissecting their mechanisms of action.

ADARreg modulates BTK inhibition by regulating ADAR2 activity

To generalize the biological relevance of ADARreg beyond the specific *in vitro* model we used for
discovery, we examined ADARreg expression in B-cell lymphoma clinical datasets in which we could
quantify the expression of ADARreg, and we found correlation with poorer overall survival in patients
with mantle cell lymphoma (MCL) and with germinal center-like diffuse large B cell lymphoma
(GCB_DLBCL) (Fig. 3A).

Therefore, we tested the ADARreg function in Jeko1, a MCL line that expresses ADARreg and shows low
sensitivity to ibrutinib. Stable ADARreg silencing significantly increased their sensitivity to ibrutinib
(Fig.3B).

To confirm that ADARreg *per se* contributed to drug resistance, we used antisense oligonucleotides to
degrade ADARreg in Jeko1 cells. Electroporation with Locked Nucleic Acids (LNAs) targeting ADARreg
reduced ADARreg expression to a level comparable to the stable ADARreg-kd polyclonal population (Fig.

S3A), and also the sensitivity to ibrutinib increased (Fig.3C). We reproduced the same phenotype in the
Toledo cell line, which is totally resistant to ibrutinib in agreement with its GCB-DLBCL origin (Fig.3D). As
further control of ADARreg-dependency of BTK inhibition resistance, we nucleofected ADARreg LNAs in
HAIRM, an ibrutinib-sensitive cell line derived from MZL or hairy cell leukemia variant. HAIRM did not
express ADARreg and did not exhibit changes in proliferation or drug sensitivity after LNA exposure
(Fig.S3B).

On the contrary, we attempted to extend the role of lncPCED1b to other lymphoma subtypes; however,
we did not observe any significant variation in BCR blockade sensitivity in either Jeko1 or Toledo cells
(Fig. S3C, D).

Given ADARreg's potential role in regulating A-to-I RNA editing [20], we studied ADARreg correlation to
ADAR enzymes. Its expression was negatively correlated with ADAR1 (Fig. S3E) and positively correlated
with ADAR2 (encoded by ADARB1) in the clinical specimens (Fig. S3F). However, ADARreg knockdown
did not significantly alter ADAR1 or ADAR2 global protein expression, neither in VL51-Ibru nor in Jeko1
(Fig.S3G).

According to ADARreg predominant nuclear localization (Fig.S3H), we hypothesized its presence could
modify ADAR enzymes' subcellular localization instead of their expression. Indeed, ADARreg stable
knockdown in VL51-Ibru showed increased ADAR2 nuclear translocation, by immunofluorescence
(Fig.3E) and by immunoblotting after nuclear/cytoplasm fractionation (Fig. 3F, S3I). ADARreg knockdown
did not affect ADAR1 subcellular distribution (Fig.S3J). We confirmed the increased ADAR2 nuclear
localization also in Jeko1 after 7 days of ADARreg LNA exposure (Fig. 3G).

As confirmation of the recovered dependency on the BCR pathway of lymphoma cells after ADARreg
silencing, we measured increased nuclear translocation of NF-κB by immunofluorescence (Fig. 3H) and
by immunoblotting (Fig. 3I, S3I). Consistently, the phosphorylation of full-length p65 increased in
resistant cells after BCR stimulation following ADARreg, but not after lncPCED1b knockdown (Fig.S3K).

ADARreg silencing affects A-to-I editing in intronic regions

To assess changes in ADAR processivity, we compared transcript variants between parental or ibrutinib-
resistant cells expressing KLK12 pgRNAs, and Ibrutinib-resistant cells with stable knockdowns of
ADARreg or lncPCED1b versus control. Using total RNA-Seq from polyclonal cell lines and the RED-ML
algorithm [28], we identified approximately 24,000 editing sites within and 13,000 outside Alu sequences
across all conditions (Table S4).

Focusing on sequence-specific editing sites detected in at least three samples, 66% (n=1,857) mapped
outside Alu sequences and 33% (n=952) within (Fig.4A, left). Differential editing analysis identified 195
non-Alu and 86 Alu-specific editing sites altered in the resistant cells. Comparable analyses of resistant
cells with ADARreg knockdown (85 non-Alu; 14 Alu) or lncPCED1b knockdown (69 non-Alu; 18 Alu)
showed that elncRNA silencing predominantly affected editing at non-Alu sites (Fig.4A, right).

We next examined the genomic distribution of editing sites. Approximately 30% localized to 3'UTR, 20%
to exons (mostly synonymous variants), 15% to exonic missense positions, and 17% to introns (Fig.4B).
ADARreg and lncPCED1b knockdown selectively altered intronic editing (Fig.4B).

In Ibrutinib-resistant cells, intronic editing was largely lost compared to parental cells (Fig.4C, left).
ADARreg knockdown reversed this trend, increasing intronic editing in protein-coding genes while
decreasing it in noncoding genes (Fig.4C, middle). lncPCED1b knockdown produced a similar increase in
intronic editing (Fig.4C, right). Many of the editing sites restored by ADARreg knockdown corresponded
to editing events originally present in parental cells but lost in resistant cells (Fig.4D). Transcripts
containing differentially edited sites were enriched for immune- and cancer-related pathways (Fig.S4A).

Altogether, these findings support a role for ADARreg, and partially lncPCED1b, in redirecting not-Alu
RNA editing towards introns of specific transcripts in BCR-independent lymphoma cells.

ADARreg affects the editing of ncRNAs involved in ibrutinib sensitivity

We wondered if the observed RNA editing changes were specifically associated with ADARreg and if the
activity was mediated by other lncRNAs, like lncPCED1b. Several elncRNAs and lincRNAs were
differentially edited in resistant versus parental cells (Fig.4E). Notably, 37 sites showed significant
differential editing upon ADARreg loss, most of which were more edited than the controls (Fig.4E).
lncPCED1b (RP11-96H19.1) was the third most edited ncRNA in ADARreg knockdown cells (Fig.4E), and
its knockdown partially restored the parental editome in resistant cells (Fig.4F). To further validate this,
we analyzed parental cells with stable knockdowns of lncPCED1b or LINC00511, a known oncogene not
associated with RNA editing. Editing sites modified upon LINC00511 knockdown showed a weaker
correlation with resistance-associated sites than those affected by ADARreg knockdown (Fig. S4B–D). In
contrast, lncPCED1b knockdown did not comparably reshape the editome of VL51 cells, in which
ADARreg did not affect ibrutinib resistance (Fig.S4E). Furthermore, intronic editing events after
lncPCED1b or LINC00511 knockdown were less pronounced in parental cells than in resistant ones
(Fig.S4F-G).

ADARreg and lncPCED1b maintain the BCR inactivation by altering RNA homeostasis

Transcriptomic analysis of cells after ADARreg or lncPCED1b knockdown revealed significant enrichment
of genesets related to the proteasome, spliceosome, and RNA degradation among the transcripts more
expressed in resistant compared to parental cells (Table S5, Fig.5A). Inhibiting these elncRNAs, we also
reversed genesets normally suppressed by BTK inhibition, in resistant cells, confirming the reactivation
of BCR dependency (Fig.5A). ADARreg and lncPCED1b inhibition reverted the SMZL signature associated
with ibrutinib-resistance [21], confirming their role in maintaining the resistant phenotype (Fig.5A).
Furthermore, resistant cells lost the differentiation features of MZ B cells (downregulation of genes
upregulated in nodal MZL vs lymph node), typically driven by BCR/NF-kB signaling pathway, adhesion
and interaction genes, like integrins and CD82, and cytokine receptors or antigen-presenting molecules,
like CD74. ADARreg and lncPCED1b knockdown markedly reconstituted their dependence on
microenvironment-driven signaling (Fig.5A). For instance, resistant cells had lower expression of genes
involved in interferon response compared to parental cells. Still, ADARreg and lncPCED1b silencing
reactivated immune responses through PRR signaling (Fig.5A, Table S6). Specifically, ADARreg
knockdown restored the expression of interferon-stimulated genes (ISGs) such as IFIH1 and DHX58,
critical in the RIG-I-like receptor (RLR) pathway that induces interferon production after detecting
unedited dsRNA, normally suppressed by ADAR editing [29] (Fig.5B).

ADARreg modulates IRF3 activity

Since ADARreg and lncPCED1b knockdown repressed many enzymes involved in RNA degradation (Fig.
5A), we assessed post-transcriptional modulation of transcript abundance using exon-intron split
analysis (EISA). Cells with ADARreg or lncPCED1b knockdown exhibited increased transcript stability,
particularly ADARreg-kd cells (106 stabilized transcripts compared to 64 for lncPCED1b) (Fig.5C, Fig.S5B,
Table S7). The stabilized transcripts were involved in metabolic pathways and response to viral infection
(Fig. 5D), while lncPCED1b knockdown primarily stabilized transcripts encoding proteasome components
(Fig.S5B). IRF3, a key transcription factor in immune response, was post-transcriptionally regulated by
ADARreg (Fig.5C). ADARreg-kd stabilized the active IRF3 isoform (IRF3-203), while destabilizing the
dominant-negative IRF3a (IRF3-202) (Fig.5E, S5C) [30]. This shift in isoform balance was confirmed at the
protein level (Fig.5F, S5D), and immunofluorescence demonstrated increased nuclear localization of IRF3
after ADARreg knockdown (Fig.5G). In resistant cells, ADARreg expression reduced all IRF3 transcripts,
with the active IRF3-203 isoform completely depleted (Fig.5E, S5C). After ADARreg knockdown, IRF3
target genes, typically downregulated in resistant cells, were partially reactivated (Fig.S5E).

ADARreg alters transcript usage

Given the massive alteration of the splicing machinery after elncRNA knockdown (Fig.5A), we next
investigated whether ADARreg depletion broadly affected transcript usage. Transcriptome analysis
showed that ADARreg knockdown induced widespread alternative transcript usage, affecting 1,536
genes, while lncPCED1b knockdown only 469 genes (Fig.S6A, Table S8). Many of these isoform changes
occurred in genes with intronic editing, with ADARreg affecting 160 and lncPCED1b 63 genes (Fig.S6A).
Variations in 3'UTR editing were rarely associated with isoform changes (Fig.S6A). Some genes with
sequence-specific editing, ENO1, RPS3, CLEC2D, MFN1, and AUTS2, displayed isoform switching after
ADARreg knockdown, with RPS3 similarly affected by lncPCED1b knockdown (Fig.S6A).

To elucidate how ADARreg regulates isoform abundance and how this is linked to intronic editing, we
performed direct RNA sequencing (DRS). This long-read sequencing approach can capture native, full-
length RNA molecules, including their modifications [31]. We precisely quantified isoform switching in
ADARreg-kd versus parental cells. Nuclear and cytosolic fractions were analyzed separately to
distinguish ADARreg effects on nuclear regulatory RNAs (such as elncRNAs) from those on mature
cytosolic transcripts. Only full-length reads were considered for downstream analyses.

In both nuclear and cytosolic compartments, a large fraction of genes produced multiple isoforms,
30.4% in the nucleus and 40.8% in the cytosol (Fig.6A). In the nucleus, the highest fraction of genes
(41.3%) transcribed noncoding isoforms, likely with regulatory function, while in the cytosol, almost the
30% produced multiple protein isoforms, indicating broad functional diversity (Fig.6A). Among nuclear
noncoding transcripts, retained intron isoforms were the most abundant, followed by processed
pseudogenes and lincRNAs (Fig.6B, top). Retained intron transcripts originated from both coding and
noncoding genes, with a similar frequency, but the latter produced generally longer transcripts (Fig.S6B).
A minor fraction of the noncoding transcripts was exported to the cytosol (Fig.6B, bottom).

ADARreg knockdown markedly altered the abundance of nuclear transcripts (Fig.S6C-D), primarily of
ncRNA (Fig.6C, top), and affected approximately 15% of protein-coding transcripts, consistently in both
the nucleus and the cytosol (Fig.6C, bottom). Among 100 full-length elncRNAs detected in the nucleus,

several were downregulated by ADARreg knockdown (Fig.6D). In the cytosol, 161 elncRNAs, were
identified, equally up- and downregulated, suggesting indirect effects (Fig.S6E).

Differential isoform usage (DIU) was quantified by TappAS algorithm [32], comparing global expression fold
changes with isoform redistribution. Globally, we noticed stronger DIU in nuclear than in cytosolic
transcripts (Fig.6E, S6F), although some genes, such as BAG1, displayed clear isoform switches at the
mature mRNA level (Fig.S6G). Nuclear-cytosolic comparisons suggested that ADARreg also regulated
mRNA export. For example, CDKN2AIP expression decreased globally after ADARreg knockdown, but
one truncated isoform (J3KNE1) was retained in the nucleus, while the main isoform (Q9NXV6), although
downregulated, remained predominant in the cytoplasm (Fig.S6H).

Among 2,272 genes, corresponding to 5,293 protein-coding isoforms detected in the nucleus, 140 were
differentially expressed. Seventy-five % of them (n=96) transcribed at least one retained intron
transcript (Fig.6E, Table S9). They were enriched for genes coding for polyA RNA binding proteins (n=25)
(Fig.6F, S6I), notably ALKBH5 and SRSF3 (Fig.6G). ALKBH5 encodes a m⁶A eraser involved in BCR signaling
[33] and in the tumor microenvironment [34], whereas SRSF3 is a splicing factor, interactor of m⁶A reader
YTHD1 [35] and already reported to regulate lncRNA splicing [36]. Both genes exhibited nuclear retained-
intron isoforms (Fig.6G). The nuclear accumulation of the ALKBH5 retained intron isoform coincided
with augmented cytosolic export of the protein-coding form (Fig.6G). These data suggest that ADARreg
modulates the nuclear abundance of retained intron transcripts, thereby post-transcriptionally
regulating genes involved in RNA homeostasis.

ADARreg regulates the editing of nuclear intron-retaining transcripts

To uncover how ADARreg contributes to RNA homeostasis, we analyzed the RNA modification events at
the isoform level. We computed the A-to-I editing variation in the samples, linking the inosine
modification to a specific RNA isoform only if supported by full-length reads. 49% of the nuclear editing
events occurred in protein-coding isoforms, 26% in processed pseudogenes, 7% in retained introns and
6% in lncRNAs (Fig.6H, left). These percentages partially reflected the abundance of each biotype.
However, using Cohen's index (ESh) to quantify effect size, ADARreg knockdown showed mild
(0.2<|ESh|<0.5) or even strong (|ESh|>0.5) impact on A-to-I editing, particularly in lncRNAs (e.g. NEAT1)
and retained intron transcripts (e.g. SRSF3) (Fig.6H, right, Fig.6I-J). Thus, editing changes within introns,
previously described in polyclonal elncRNA-silenced cells (Fig.4C), primarily reflected altered nuclear
editing of retained-intron transcripts (Fig.6I-J, top).

In contrast, cytoplasmic editing events were almost exclusively in protein-coding isoforms, though the
few lncRNA-associated sites (0.6%) were strongly ADARreg-dependent (Fig.S6J).

ADARreg regulates RNA homeostasis also by means of m⁶A RNA methylation

The modulation of nuclear abundance of retained intron transcripts and mRNAs encoded by ALKBH5
and SRSF3, or ENO1 and PKM, well-recognized m⁶A targets[37,38], suggested that ADARreg might also
influence RNA methylation. Applying isoform-resolved m⁶A quantification, we found that 35% of nuclear
m⁶A peaks occurred in processed pseudogenes (Fig.6H), consistent with the well-documented role for

m⁶A-coupled decay of those transcripts [39]. ADARreg knockdown significantly altered m⁶A abundance in
30-40% of sites across major transcript biotypes, including protein-coding, pseudogene, retained-intron,
and lncRNA isoforms (Fig.6H). Unlike A-to-I editing, strongly m⁶A-regulated sites were similarly
distributed across transcript classes (Fig.6I, right). Consistent with increased cytosolic ALKBH5 mRNA,
overall nuclear m⁶A levels tended to decrease upon ADARreg knockdown (Fig.6I–J).

We investigated whether variations in A-to-I editing could affect the m⁶A abundance at specific targets.
Comparing nuclear transcripts with exclusively m⁶A modulation and transcripts where both m⁶A and A-
to-I significantly changed after ADARreg knockdown revealed that both marks were significantly
different in protein-coding genes and pseudogenes (Fig.S6K, 6K). The increased A-to-I events in
pseudogenes were associated with an increase in m⁶A (Fig.6K, right). To investigate the role of RNA
editing in the context of nuclear pseudogenes, we compared the differential expression of cytosolic
isoforms of paired parent genes, specifically when the pseudogene was modulated in terms of m⁶A and
A-to-I editing, or m⁶A only. Parent genes of edited pseudogenes displayed greater isoform variability
(logFC variance mean, μ = 0.34) than those paired with only methylated pseudogenes (logFC variance
mean, μ = 0.146) (Fig.6L). Parent genes paired with edited pseudogenes were primarily involved in RNA
homeostasis, controlling mRNA metabolic processes and localization, specifically NPM1 and HNRNPD
(Fig. 6M).

For nuclear protein-coding transcripts bearing both modifications, increased editing was mainly
associated with reduced m⁶A. These transcripts were predominantly mono-exonic (e.g., histone genes),
typically more methylated than spliced RNAs [40,41]. The A-to-I editing in these transcripts could induce
their demethylation and influence stability (Fig. S6K).

ADARreg affects protein-coding isoforms involved in immune regulation

To investigate the effects of the observed transcriptional and post-transcriptional changes, we
performed a functional enrichment analysis of the protein-coding isoforms expressed in the cytosol,
using the TappAS algorithm [32]. We identified four clusters of enriched gene ontology terms: cluster 1,
related to chemokine-mediated signaling pathway (e.g., CCR5 chemokine receptor binding); cluster 2
and 3, related to the negative regulation of immune effector process and of interferon-gamma
production; cluster 4, related to the positive regulation of calcium ion transport (Fig.7A). Key
upregulated transcripts included CD38, a marker of BCR activation [42], and XBP1, which regulates plasma
cell differentiation [43]. Conversely, TNFSF4, CCL3, CCL4, CCL5, XCL1, and IL23A were downregulated,
indicating a potentially impaired recruitment of a protective immune niche [44]. Reduced CEACAM1, which
modulates BCR signaling [45] and NK cell activity [46], further supports immune suppression by ADARreg.

To investigate the direct transcriptional effect of ADARreg knockdown, we correlated the modulation of
expression and methylation of each isoform after ADARreg silencing. Coherent with the observations in
the nucleus, we found that the most significantly modulated isoforms lost m⁶A marks and were
upregulated, including the isoforms of the genes coding for the ribosomal protein S3 (RPS3) and CLEC2D,
the ligand for the NK/T-cell receptor CD161 [47,48] (Fig.7B), which had already been robustly characterized
by an alternative transcriptional mechanism and intronic editing (Fig. S6A).

The most downregulated transcripts, encoding CCL5 and XCL1, were enriched for m^{6}A marks after
ADARreg knockdown, consistent with enhanced transcript decay (Fig.7B).

DRS also validated IRF3 isoform modulation: the longer, active isoform was stabilized after ADARreg
knockdown, likely via competition with a retained-intron isoform sharing its 3⍰ UTR miRNA sites (Fig.
S7A, Fig.4C). These findings support that the effects on RNA stability due to ADARreg depletion are likely
mediated by a precise deposition of m^{6}A marks on selected isoforms in an ADARreg-dependent manner.

ADARreg modulates the RPS3 and NF-κB interaction

To demonstrate the impact of the ADARreg-regulated isoform on BCR blockade in lymphoma cells, we
investigate the role of *rpS3*. The latter is a multifunctional protein involved in translation, DNA repair,
apoptosis, and immune regulation, by interaction with p65 to activate specific genes [49]. RPS3 has
different domains responsible for apoptosis and DNA damage (Fig.7C)[50]. ADARreg and lncPCED1b
knockdown restored expression of the RPS3-208 isoform (ENST00000527273), which was low in
resistant cells (Fig. S7B). This isoform lacks the DNA repair domain but retains the apoptosis-inducing
and p65-interacting domains (Fig.7C, left).

Coimmunoprecipitation confirmed that ADARreg or lncPCED1b depletion increased N-terminal
truncated rps3 and the interaction of rps3 with the phosphorylated form of p65 in the nucleus (Fig. 7D).
DRS confirmed multiple RPS3 isoforms, including the truncated variant retaining the p65-binding domain
and harboring a long 3⍰ UTR derived from intron 4–5 retention (Fig. 7C, right). Altogether, these data
suggest that RPS3 isoform regulation is a key mechanism linking ADARreg to NF-κB-driven immune
reactivation.

ADARreg inhibits NK cell-mediated cytotoxicity

Given the modulation of the interferon response through IRF3, the RPS3-driven p65 activity, and the
chemokine-related pathways, we hypothesized that ADARreg silencing could enhance lymphoma
immunogenicity. ADARreg knockdown reduced CEACAM1 and altered CLEC2D (LLT1) isoform usage (Fig.
S7C, left). DRS detected multiple CLEC2D isoforms (Fig. S7C, right) with ADARreg knockdown favoring a
shorter, less methylated isoform (Fig. 7B, E, S7D), suggesting functional modulation of NK-cell
interaction.

Co-culturing ADARreg knockdown VL51 cells with NK cells from a healthy donor resulted in significantly
higher NK cell-mediated killing compared to control cells (Fig. 7F, S8A). Additionally, NK cells survived
longer in co-cultures with ADARreg knockdown cells (Fig. S8B).

To assess clinical relevance, we analyzed single-cell RNA-seq data from chronic lymphocytic leukemia
(CLL) patients undergoing ibrutinib therapy [51]. Using genes altered by stable ADARreg-kd as a surrogate
for ADARreg level (undetectable in polyA datasets), we defined an ADARreg^{low}-score to compare CLL
cells from two fast and two slow responders (Fig.S8C). Transcriptomic profiles of surviving cells after
treatment resembled the SMZL ADARreg^{high}- VL51-Ibru model (Fig. S8C). Similar cells were also more
abundant at the pretreatment time point in slow responders (Fig.S8C). We analyzed NK cells in

untreated samples, annotating them as cytotoxic, inflammatory, or tumor-associated NK (TANK)-cells (Fig.S8D) [52], and the tumor cells, on the basis of ADARreg signature, as ADARreg^high^-like or ADARreg^low^-like B cells. Therefore, we analyzed cell-cell communication using CellChat (Fig.7G,S8E). At pretreatment, in a slow responder, ADARreg^high^-like B cells increased CLEC2D-CD161 interactions with active NKs, compared to ADARreg^low^-like B cells (Fig.7G). Conversely, ADARreg^low^-like B-cells interacted more with TANK-cells through CD99-PILRA signaling, recently associated with favorable immunotherapy outcomes [53] (Fig.7G). No differences in cell communication between ADARreg ^high/low^-like B- and NK-cells were significant in the fast responder (Fig.S8E). Tumor cells persisting after treatment had higher CLEC2D expression and ADARreg signature in both patients (Fig.7H). However, those cells were more abundant in slow than in fast responders (Fig.7H). In summary, ADARreg modulated CLEC2D isoform balance, favoring NK inhibition in ibrutinib-resistant cells in vitro. A transcriptional program, like the one controlled by ADARreg, linked high CLEC2D expression to ibrutinib resistance in CLL patients. Finally, the tumor cells resembling the ADARreg ^high-^immune-silent VL51 interact with protumoral NK-cells through CLEC2D.

Discussion

This study demonstrated that LOC730338, which we renamed ADARreg, plays a key role in maintaining immune suppression and refractoriness to BCR signaling regulation in lymphoma cells. We first identified its role in modulating sensitivity to the BTK inhibitor ibrutinib in an *in vitro* model of MZL with secondary BTK inhibitor resistance and subsequently extended these findings to MCL and GCB-DLBCL models, as well as CLL patients. Collectively, our results link ADARreg to the control of the inflammatory response through a complex regulatory network, based on RNA homeostasis (Fig.S9).

Although ADARreg was previously associated with RNA editing, its function remained unknown [20]. The connection between RNA editing and immune regulation is well established: upon pathogen recognition, PRRs activate pathways such as NF-κB, MAPK, and type I interferon [54,55], which must be tightly controlled to prevent excessive activation[56]. ADAR1 and ADAR2 maintain immune tolerance through A-to-I RNA editing of self-dsRNA [57] and by modulating the transcript stability and alternative splicing of genes involved in the antiviral response [58]. Since A-to-I editing can drastically alter RNA's structural and coding potential [59,60] perturbation of this process has profound effects. Indeed, ADARreg knockdown disrupted RNA degradation and splicing machineries, highlighting its importance in maintaining RNA homeostasis.

Mechanistically, ADARreg regulates the nuclear translocation of ADAR2, a less abundant but more sequence-specific editor than ADAR1. An initial analysis of RNA editing on transcriptomic data revealed editing alteration upon ADARreg knockdown, particularly outside repetitive Alu sequences, consistent with an ADAR2-dependent mechanism [29,61]. Notably, ibrutinib-resistant cells displayed a distinct RNA editing pattern that was reversed by ADARreg knockdown. Editing changes also affected noncoding transcripts, including elncRNAs such as lncPCED1b, suggesting a link between ADARreg and epigenetic adaptation to drug resistance. Indeed, we observed altered SEs activation and elncRNA transcription in cells with acquired refractoriness to BCR activation, implicating ADARreg and lncPCED1b in the BCR pathway regulation.

Our findings suggest that ADARreg operates within a broader network of ncRNAs that are modulated by
ADARreg itself and are subject to RNA modifications. Some ncRNAs, such as lncPCED1b, appear to
mediate ADARreg's transcriptomic effects. We propose that ADARreg, controlling ADAR2 localization
and eIncRNA expression, regulates the stability and localization of protumoral transcriptional isoforms.
Consistent with this, ADAR2 has previously been reported as a potential tumor suppressor in glioma [62],
and alternative splicing frequently promotes pro-tumorigenic isoforms [60,63].

Although globally informative and practical, RNA editing estimation on total RNA-seq cannot fully
resolve how A-to-I editing affects isoform-specific splicing, as multiple transcripts may overlap the same
editing sites. Recent advances in chemistry and analytical tools have made DRS a powerful technology to
study isoform switching and isoform-resolved RNA modifications in subcellular compartments [31,64,65].
Using this strategy and leveraging our adapted *in silico* pipeline to enhance the single-molecule
resolution of the method, we found that ADARreg regulates the A-to-I editing of nuclear-retained intron
transcripts and influences the alternative expression of genes involved in RNA homeostasis. These
include multiple RNA-binding proteins that regulate RNA trafficking, maturation, and post-
transcriptional modifications, like the m^6A eraser ALKBH5 and the SRSF3, interactor of the m^6A reader
YTHDC1 [66]. Accordingly, ADARreg knockdown reduced nuclear m^6A levels, consistent with the
downregulation of ALKBH5. Interestingly, inosine modifications appeared to protect specific
pseudogenes from RNA demethylation, enhancing their nuclear retention or stability. The modulation of
pseudogenes, which sequester miRNAs and other binding factors in the cytosol, exposes parent genes to
altered post-transcriptional regulation of their isoforms. One key target of this mechanism, NPM1,
encodes nucleophosmin 1, a nucleolar protein that forms the NPM–ALK fusion in anaplastic large cell
lymphoma and is deregulated in up to 30% of acute myeloid leukemias through exon 12 mutations that
cause its aberrant cytoplasmic mislocalization [67]. Given its nucleolar localization and shuttling activity,
NPM1 may also influence ADAR2 trafficking; however, a direct interaction remains unproven. Notably,
NPM1 also acts as a chaperone for NF-κB [68] and promotes immune evasion in several cancers [69].

The modulation of RNA nuclear retention is also linked to ADARreg through its impact on NEAT1, a
nuclear lncRNA with scaffold function in organizing paraspeckles [70]. Those nuclear bodies are required
for the storage of proteins, like transcription factors, and RNA species, like A-to-I edited mRNAs and pri-
miRNAs, as a survival strategy to prepare the cell for stress [71]. NEAT1 is also a documented target of the
ALKBH5 enzyme, which stabilizes it to modulate the availability of repressors and the expression of
immune-stimulating chemokines [72]. Several observations suggest that NEAT1 and paraspeckles can
influence the innate immune response, although the complete mechanism remains to be fully
elucidated [73].

Overall, our data suggest that ADARreg influences immune signaling by modulating the BCR receptor
and the downstream NF-κB signaling pathway at multiple levels. Like the reported interaction between
ALKBH5 and a lncRNA (treRNA1) that modulates key transcripts, including those involved in BCR
signaling in DLBCL [33], ADARreg appears to drive a comparable mechanism. Among its downstream
targets, RPS3 was the most demethylated and upregulated cytosolic transcript following ADARreg
silencing. RPS3, beyond its ribosomal functions, is involved in apoptosis, DNA repair, and transcriptional
regulation [74]. It was identified as a non-Rel subunit of NF-κB responsible for activating its immunogenic
signature [49,75]. We identified two RPS3 isoforms, with a p65-interacting domain, that enhance nuclear

binding of phosphorylated NF-κB upon knockdown of ADARreg or its downstream lncRNA target, lncPCED1b. Finally, BCR and cytokine receptor signaling were reactivated, restoring sensitivity to BCR blockade.

ADARreg also contributes to immune evasion by inhibiting interferon response and favoring the expression of pro-tumoral cytokines. The active IRF3 isoform is post-transcriptionally downregulated by ADARreg, which induces the expression of a retained-intron transcript that competes with the main isoform for miRNA binding. IRF3 is a downstream target of the TLR pathway triggered by pathogen infection and is crucial for the induction of interferon-stimulated genes, as well as the subsequent innate response. Consequently, ADARreg knockdown enhanced innate immune activation and increased SMZL cell susceptibility to NK cell cytotoxity by modulating LLT1 (CLEC2D). Notably, CLEC2D expression correlated with m^6A-dependent isoform changes. Conversely, pro-tumoral cytokines such as CCL5 and XCL1 were destabilized due to increased m^6A, demonstrating that ADARreg does not globally alter methylation but fine-tunes epitranscriptomic marks to restore immune competence. In SMZL patients, poorer outcomes are associated with an immunosuppressive phenotype with immune checkpoint activation [76]. These patients often exhibit KLF2 and NOTCH2 mutations, but a clear mechanism for immune escape is missing. ADARreg may be a critical factor influencing tumor-microenvironment interactions, as sustained by the worse outcome in ADARreg[high]-GCB-DLBCL patients, who predominantly exhibit "immune-cold" microenvironment signatures [77]. This mechanism was further validated in CLL patients, where we observed a link between the selection of tumor clones resistant to BCR inhibition and the suppression of an immunological signature modulated by ADARreg. Moreover, an ADARreg[high]-signature in hematological tumors, such as MZL, CLL, and MCL, may indicate an unfavorable prognosis for patients treated with BCR signaling inhibitors. Conversely, ADARreg represents a promising therapeutic target for antisense oligonucleotide strategies to elicit spontaneous immune responses against immune-suppressive cells.

Limitations of the study

Although our results clearly associated ADARreg with the hijacking of a complex regulatory mechanism of RNA processing and with the subversion of the tumor-microenvironment immune crosstalk, we have not yet identified how ADARreg modulates all this in the first place. The broad molecular events conditioned by ADARreg knockdown imply that it is a highly upstream node in the network that directs the epigenetic remodeling of immune-silent lymphomas. Furthermore, additional experiments will be necessary to identify the direct RNA or protein interactors of ADARreg.

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Author contributions

L.C.: conceptualization, analysis, visualization, and manuscript preparation; F.G.: investigation and
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investigation; S.Z.: analysis; N.M.: investigation; S.A.: investigation. A.J.A.: investigation and analysis;
R.G.: investigation and resources. R.J.: conceptualization and resources; F.B.: conceptualization and
resources; S.N.: conceptualization, investigation, analysis, visualization, manuscript preparation, project
supervision and resources.

Disclosure of Potential Conflicts of Interest

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nothing to disclose.

 Main Figures

Figure 1. Ibrutinib-resistance is associated with changes in super-enhancer activation

(A) ChIP-seq H3k27ac peak counts in biological replicates of parental VL51 and VL51-Ibru (IBRr). Total
peaks (horizontal bars) and robust differentially enriched peaks in VL51 and VL51-Ibru are shown
(vertical bars, red and green, respectively). The indicated enhancer numbers consider H3K4me1 peak
presence and H3K4me3 absence. **(B)** Example of an enhancer lost in VL51-Ibru near CDKN2A/B. Tracks
show H3K4me1, H3K4me3, and input from VL51 (green) and VL51-Ibru (red), with H3K27ac peaks and
differentially expressed transcripts highlighted. **(C)** Volcano plots of genes within 40 kb of enhancers lost

(left) or gained (right) in VL51-Ibru cells. Orange and green denote significantly down- and upregulated
575 genes; selected genes are labeled.

Figure 2. Functional screening of eIncRNAs linked to BCR blockade

(A) CRISPRi strategy scheme. **(B)** Composition of the CRISPRi library. **(C)** Workflow of the CRISPRi screen.

(D) MDS plots of samples at baseline (T0) or after 6, 10, and 14 days. VL51 (green) and VL51-Ibru (red)
exposed to ibrutinib (triangles) or vehicle (circles) are shown. VL51-Ibru treated with ibrutinib cluster
together. **(E)** Scatter plots of β-values reflecting dropout of lincRNAs (left) or eIncRNAs (right) in VL51

 versus VL51-Ibru after 14 days of ibrutinib. Statistically significant hits (FDR < 0.2) are labeled; selected
583   candidates are highlighted. **(F)** Expression of ADARreg and lncPCED1b, and ibrutinib sensitivity of
584   monoclonal knockdown lines compared with KLK12 controls (n=3,* P <0.05).

Figure 3. ADARreg modulates BTK inhibition response by regulating ADAR2 activity

(A) Kaplan–Meier survival of MCL (n= 102) or GCB-DLBCL (n= 172) patients stratified by high or low ADARreg expression. **(B)** Growth curves of Jeko1 KLK12 or ADARreg-kd cells under vehicle or ibrutinib. **(C and D)** Growth curve of Jeko1 (C) or Toledo (D) exposed to ibrutinib or vehicle, 4 days after ADARreg LNA or control electroporation. **(E)** ADAR2 immunofluorescence in parental VL51, ibrutinib-resistant KLK12-kd, ADARreg-kd, or lncPCED1b-kd cells. Representative images (left) and nuclear/cytoplasmic quantification of ADAR2 integrated density (right). DAPI-stained nuclei, blue, Alexa647-conjugated antibody-stained ADAR2, red. Close-up from each picture shows a single-cell level magnification. **(F)** Nuclear/cytosolic ADAR2 protein levels in ibrutinib-resistant KLK12-kd and ADARreg-kd VL51 clones. Average of three independent monoclonal cell lines per group. **(G)** Nuclear/cytosolic ADAR2 protein levels in Jeko1 7 days after ADARreg-LNA or controls, left, and relative quantification, right. H3 was used to normalize nuclear lysate and vinculin, the cytosolic. **(H)** NF-κB immunofluorescence in parental VL51, ibrutinib-resistant KLK12-kd, ADARreg-kd, or lncPCED1b-kd cells. Representative images (left) and nuclear/cytoplasmic quantification of NF-κB integrated density (right). DAPI-stained nuclei, blue, Alexa647-conjugated antibody-stained ADAR2, red. Close-up from each picture shows a single-cell level magnification. **(I)** NF-κB p65 nuclear/cytosolic protein levels across three monoclonal lines per group. (n=3, P.value<0.05)

Figure 4. ADARreg silencing alters A-to-I editing in introns.

(A) Editing sites inside Alu (red) or non-Alu (blue) regions (left), and significant differentially edited sites in ibrutinib-resistant versus parental, or ibrutinib-resistant ADARreg-kd or lncPCED1b-kd versus controls (right)(p-value < 0.05, Limma test). **(B)** Genomic distribution of total and differentially edited sites in

ibrutinib-resistant vs parental, or ibrutinib-resistant ADARreg-kd or lncPCED1b-kd vs controls (p-value <
0.05, Limma test). **(C)** LogFC of non-Alu editing sites across genomic categories in ibrutinib-resistant
versus parental (left) and in ADARreg-kd (middle) or lncPCED1b-kd (right) versus control (pvalue< 0.05,
Limma test, mean and outlier values are shown). **(D)** Correlation of editing site logFC in ibrutinib-
resistant vs parental and ADARreg-kd vs control; significant sites are highlighted. The regression line
represents Pearson correlation (p-value < 0.05, Limma test). **(E)** Correlation of ncRNA editing site logFC
in ADARreg-kd vs control and ibrutinib-resistant vs parental; significantly affected sites (blue) or those
shared with lncPCED1b-kd (red circled) are marked (p-value < 0.05, Limma test). **(F)** Correlation of
editing site logFC in ibrutinib-resistant versus parental and lncPCED1b-kd versus control; significant sites
are highlighted. The regression line represents Pearson correlation (p-value < 0.05, Limma test).

Figure 5. ADARreg modulates IRF3 activity

(A) GSEA of ibrutinib-resistant versus parental, ibrutinib-resistant ADARreg-kd or lncPCED1b-kd versus control. Normalized enrichment score and significance are represented. **(B)** Scatter plot of interferon-stimulated gene logFC in ibrutinib-resistant vs parental and ibrutinib-resistant ADARreg-kd vs control. **(C)** Changes in exonic (Δexon) versus intronic (Δintron) reads per transcript in ibrutinib-resistant ADARreg-kd vs control; post-transcriptionally modulated transcripts (Δexon/Δintron≠1) are highlighted (P-

value<0.05). **(D)** gProfiler gene ontology analysis of post-transcriptionally modulated transcripts. **(E)** IRF3

isoform abundance in parental, ibrutinib-resistant control, and ibrutinib-resistant ADARreg-kd cells

(*FDR < 0.05). **(F)** IRF3/IRF3a protein quantification in ibrutinib-resistant control and ADARreg-kd cells

(n=3, * P <0.05). **(G)** Left, IRF3 immunofluorescence in parental, ibrutinib-resistant KLK12-kd, ADARreg-

kd, or lncPCED1b-kd cells. DAPI-stained nuclei, blue; Alexa647-conjugated antibody-stained IRF3, red.

Close-up from each picture shows a single-cell-level magnification. Right, nuclei/cytoplasm ratio of IRF3

integrated density (n=3, * P <0.05).

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Figure 6. ADARreg regulates RNA homeostasis at multiple levels.

DRSeq full-length RNA profiling of nuclear and cytosolic transcripts in VL51-Ibru ADARreg-kd and
controls.

(A) Proportion of genes producing single/multiple transcripts and coding for zero/one/multiple isoforms
in the nucleus (top) or in the cytosol (bottom). **(B)** Nuclear (top) or cytosolic (bottom) transcript biotype
distribution. **(C)** Nuclear (top) or cytosolic (bottom) differentially expressed coding and noncoding
transcripts (FDR<0.05). **(D)** Volcano plot of nuclear eIncRNAs. Statistically significant hits (FDR < 0.05) are
labeled. **(E)** Top, isoform usage change versus gene expression logFC between ADARreg-kd and control
in the nucleus. Top genes with major isoform switches (blue) are labeled. Bottom, numbers of protein-
coding genes subject to DIU, involving a retained intron isoform, in the nucleus. **(F)** Ranking of isoform
usage changes in polyA-RNA-binding proteins, highlighting m^6A RNA methylation regulators. **(G)** Top,
gene, transcript, and protein-level expression of ALKBH5 and SRSF3 in ADARreg-kd and control in the
nucleus (squares, replicate expression; points, mean replicate expression). Middle, tappAS graphical
representation of transcript-level annotations. (Grey boxes, exons; dotted lines, spliced introns; black
line noncoding retained sequences). Bottom, ALKBH5, gene, transcript, and protein-level expression in
the cytosol. **(H)** A-to-I and m^6A site distribution by biotype of nuclear transcripts detected by DRS full-
length-read, left, and by effect size, in the comparison ADARreg-kd versus control (not modulated
(|ESh|<0.2), mild (0.2<|ESh|<0.5), strong (|ESh|>0.5)) **(I)** Aggregated effect sizes (ESh mean of
modulated sites by transcript) for A-to-I, left, and m^6A, right, across transcript biotypes. **(J)** Proportion of
A-to-I (top) and m^6A (bottom) significantly modulated sites, aggregated by transcript, in control and
ADARreg-kd samples. Retained intron transcripts, IncRNAs, and pseudogenes are highlighted. **(K)** Top,
distribution of nuclear transcripts carrying ADARreg-dependent m^6A ± A-to-I changes by transcript
biotype; bottom, scatter plot of A-to-I and m^6A ESh values of nuclear processed pseudogenes. **(L)**
Variance distribution of the cytosolic isoform logFC by parent gene, paired with differentially
edited/methylated nuclear pseudogenes after ADARreg-kd (blue), or paired with only differentially
methylated nuclear pseudogenes (orange). Mean, μ, and standard deviation, ⬚, of the variance are
indicated. **(M)** Semantic clustering of enriched gene ontology terms in parent genes of m^6A/A-to-I-
modulated pseudogenes

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 Figure 7. ADARreg suppresses NK-cell cytotoxicity.

(A) GO cluster analysis of differentially expressed cytosolic protein-coding transcripts in ibrutinib-
resistant ADARreg-kd vs control cells. Pie chart area represents DE regulation measured by relative -
log10(p value) of the enrichment test, and node size represents the number of genes annotated for each
GO term. **(B)** Correlation of t-statistic for ADARreg-kd versus control and transcript-aggregated m^{6}A ESh
in cytosolic protein-coding transcripts. Strongly significantly downregulated (red), upregulated (green) or
mildly modulated (yellow) transcripts are highlighted. **(C)** Left, domain structures of rps3 isoforms
encoded by RPS3-203/208. Right, TappAS graphical representation of transcript-level and protein-level
annotations of RPS3. Grey boxes, exons; dotted line, spliced introns; black line, noncoding retained
sequences; red and green boxes, protein domains. **(D)** Co-immunoprecipitation of phosphorylated p65
with long and short rps3 isoforms in ibrutinib-resistant ADARreg-kd, lncPCED1b-kd, or control cells. **(E)**
Expression of CLEC2D protein-coding isoforms by DRS in ibrutinib-resistant ADARreg-kd versus control.
(F) Viable GFP$^+$-mCherry$^+$ tumor cells after coculture with donor NK cells for 7 days, normalized to
tumor-alone controls (n=3,* P <0.05). **(G)** CellChat-derived communication networks between ADARreg-
high or -low B cells and NK subsets in an ibrutinib slow-responder CLL patient. **(H)** Single-cell expression
of CLEC2D and the ADARreg signature in B cells from fast or slow ibrutinib responders at pretreatment
and during therapy.

STAR METHODS
Cell lines, LNA transfection, and drug treatments
All cell lines were grown in RPMI supplemented with 10% heat-inactivated fetal bovine serum (FBS). The
VL51 parental and VL51-Ibru were obtained as previously described [11,78]. HEK293T cells used for
lentiviral particle production were grown in DMEM supplemented with 10% FBS. Cells were maintained
at 37°C in humidified incubators under 5% CO$_2$. All cell lines were tested negative for Mycoplasma
contamination and were verified for identity by STR profiling. LNA ADARreg1-3 and negative controls
NC1-2 were bought from IDT, Integrated DNA Technologies. Sequences are reported in Supplementary
Table S10. Tumor cells (200.000 per sample) were resuspended in 20 ml of SG solution (Amaxa-Lonza)
with LNA (1 nmol) and electroporated using 4D Nucleofector (Amaxa-Lonza), according to the
manufacturer's instructions and incubated for 96h, before further treatment or RNA extraction. Cells
were exposed to ibrutinib (Selleckchem, Houston, TX, USA) or DMSO (Sigma) for 96h.
IgG stimulation
VL51 parental or ibrutinib-resistant cells (5 million) were collected, washed with PBS, and exposed to 20
mg/ml of IgG (Goat Fab Anti-Human IgG, UNLB) for 20 min. Cells were washed and lysed for protein
extraction.
Cell proliferation assays
Response to drug treatments was evaluated at the endpoint by seeding 100,000 cells in two sister 24-
well plates, each with serial dilutions of the drug or vehicle. The mitochondrial functionality was
measured by MTT (3-(4,5-dimethylthiazol-2-yl)-2,5-diphenyl tetrazolium bromide) assay at 0 or after 96
h. Cell growth curves were measured by the Incucyte instrument. In detail, stable infected cells or
transient silenced cells were seeded 5,000 cells/ well for VL51, 10,000 for Jeko1, or 20,000 for HAIRM in
the presence of ibrutinib (5 mM for VL51-Ibru, KLK12kd or ADARreg kd; 40 nM for Jeko1, KLK12kd or

ADARreg kd, 1 mM for LNA-transfected Jeko1, or 20 nM for LNA-transfected HAIRM and followed over
time. Green fluorescence was measured for VL51, while cell-by-cell and confluence analyses were
performed for Jeko1 and HAIRM cells.

ChIP-seq

The ChIP experiment followed standard protocols with an H3K27ac (Active Motif Cat# 39133,
RRID:AB_2561016) H3H4me1 (Abcam Cat# ab8895, RRID:AB_306847) and H3K4me3 (Millipore Cat# 17-
614, RRID:AB_11212770) antibody or control IgG (Millipore Cat# 12-371, RRID:AB_145840) as negative
control. In brief, 50 million cells were resuspended and fixed with 1% methanol-free formaldehyde.
Nuclei were extracted, washed and resuspended 25 million/ml in sonication buffer (10 mM Tris pH 8.0,
100 mM NaCl, 1 mM EDTA, 0.5 mM EGTA, 0.5% N-lauroylsarcosine sodium salt and fresh 0.1% Na-
deoxycholate and proteinase inhibitors). Chromatin was fragmented by sonication with Covaris to an
average size of 200-500 bp. A 50 μL aliquot was saved as input, whereas 650 μL (corresponding to 16
millions of nuclei) each were incubated with 10 μg antibody, overnight at 4°C. DNA-protein complex was
recovered overnight at 65°C, then proteins and RNA were digested and DNA purified by PCR purification
kit (Qiagen)

Library preparation from 5 ng purified chromatin was performed using NebNext Ultra II DNA Library
Prep with magnetic purification beads (E7103S, New England Biolabs) and multiplex oligonucleotides
(Dual Index Primers; E7600S, New England Biolabs), followed by adaptor ligation according to the
manufacturer's protocol. Quality controls of amplified chromatin fragments were performed on
Bioanalyzer 2100 (Agilent Technologies) and Qubit V4 (Thermo Fisher Scientific). Next-generation
sequencing was performed on a NextSeq 2000 (Illumina) using the P2 reagent kit V3 (100 cycles;
Illumina). Samples were processed starting from stranded, single-ended 120bp-long sequencing reads.

Raw data in Fastq format were first preprocessed to ensure suitability for downstream analysis. Adapter
sequences were trimmed from the reads using trimmomatic algorithm[79]. Quality control was performed
to filter out reads with low-quality bases, short lengths, or those suspected to be sequencing artifacts.
Reads were mapped to the HG38 human genome using BWA alignment. Peak calling was performed
using MACS (version 2) with "broadPeak" setting to identify broad regions of H3K27me3, then
annotated to the nearest genes with HOMER considering both transcription start and end sites.

Raw and processed data have the accession number (pending) and were uploaded on Gene Expression
Omnibus (link pending).

Total RNAseq

Total RNA was isolated from cell lines using a phenol:chloroform extraction method. RNA samples were
treated with DNase I (Qiagen). Quality control for extract RNA was performed on the Agilent BioAnalyzer
(Agilent Technologies, Santa Clara, CA, USA) using the RNA 6000 Nano kit (Agilent Technologies, Santa
Clara, CA, USA). Concentration was determined by the Invitrogen Qubit (Thermo Fisher Scientific,
Waltham, MA, USA) using the RNA BR reagents (Thermo Fisher Scientific). The TruSeq RNA Sample Prep
Kit v2 for Illumina (Illumina, San Diego, CA, USA) was used for cDNA synthesis and adding barcode
sequences. The sequencing of the libraries was performed via a paired-end run on a NextSeq500
Illumina sequencer (Illumina). As an average, 25 million reads were collected per sample.

For the RNA-Seq analysis, the sequencing reads were first mapped to the human reference genome
(hg38) using the STAR aligner [80], ensuring high-quality alignment and accurate mapping. Gene/Isoform

expression was then quantified using RSEM, leveraging the Gencode v22 annotation as the gene model
to associate reads with genomic features effectively.
To identify differentially expressed genes (DEGs), the limma package was employed, which uses a linear
modeling framework to analyze expression data. To account for the multiple comparisons inherent in
high-throughput experiments, the Benjamini-Hochberg (BH) procedure was applied to adjust p-values,
ensuring control of the false discovery rate (FDR).

TSS selection
For each putative enhancer RNA (eRNA), the corresponding transcription start site (TSS) was determined
based on the publicly available CAGE data produced by the FANTOM consortium. This approach
associates eRNAs with experimentally validated TSS. An ad-hoc R script was developed to automate the
integration of eRNA coordinates with the CAGE TSS annotations.

CRISPRi library design
pgRNAs library was designed as previously described in [81,82] with a few modifications to better fit CRISPRi
(https://github.com/sunandinini/CRISPETa_i-a/tree/bug-fix). 99 positive controls were chosen among
validated essential genes[83] expressed in 43 of our lymphoma cell lines, while 33 neutral controls were
selected among genes completely not expressed in any of the 43 lymphoma cell lines.
6-10 independent constructs per each RNA to be tested, 4618 pgRNAs versus eRNAs and lincRNAs and
1207 control pgRNAs, for a total of 5825 constructs. All pgRNA sequences are listed in Table S2. The
custom CRISPRi Library was cloned and expanded in outsourcing (Vector Builder Inc.)

Stable dCas9-ZIM3-expressing cells preparation
VL51 parental or ibrutinib-resistant were infected with the lentiviral vector pHR-UCOE-SFFV-dCas9-
mCherry-ZIM3-KRAB (RRID:Addgene_154473). Lentivirus production was carried out by co-transfecting
HEK293T cells with 6 mg of dCas9 plasmid, 4.5 mg psPAX2 plasmid, and 1.5 mg of the packaging pVsVg
plasmids, using JetPrime (Polyplus). 24 h before the transfection, 3 million HEK293T cells were seeded in
a 10 cm dishes. The supernatant containing viral particles was harvested 48, and 72 h after transfection.
Viral particles were then concentrated 100-fold by adding 1 volume of Lenti-X Concentrator (Takara) to
every three volumes of supernatant. After 12 h at 4 C, the supernatant was centrifuged at 1,500 g for 45
min at 4 C, resuspended in complete medium, and stored at -80 C until use.
For the generation of stable dCas9-expressing cell lines, cells were incubated for 72 h with medium
containing concentrated viral preparation and 8 mg/mL polybrene. MCherry+ cells were sorted several
times before obtaining cell lines able to sustain dCas9-ZIM3-KRAB expression and cultured in the
presence of additional nonessential amino acids and higher content of fetal bovine serum. Cells were
also tested for dCas9 expression by western blotting, and the protein's functionality was proven by
infection with sgRNA directed against MALAT1 TSS (pDECKO_Malat1_C, RRID: Addgene_72622). Two
independent dCas9 clones for parental and two for resistant cells were selected as biological replicates
to infect in parallel during the screen.

Lentiviral library titration
The functional titer of the CRISPRi library was evaluated, infecting a small scale of dCas9-ZIM3-KRAB-
mCherry cells with an increasing amount of virus. After 48 and 72 hours, cells were checked for GFP
expression by FACS, and the volume of virus per million cells was calculated based on the infection that
achieved 30% GFP+ cells.

elncRNA CRISPRi screen

Fifty million cells were infected with the CRISPRi Library at an MOI of 0.3 and then selected by puromycin exposure for two days, followed by a two-day washout period. One-third of the cells, corresponding to a minimum of 3 million, were saved as T0 to obtain the baseline representation of the pgRNAs distribution. Then, 3 million mutant cells per experiment group were selected by exposure to ibrutinib or DMSO and culture under drug pressure for 14 days. The number of cells was calculated to represent 500 cells per each of the 6000 pgRNAs. The dose chosen for the treatment was 500 nM of ibrutinib (a sub-active dose equivalent to 1/10 of the IC50) in parental cells. In parallel, mutant cells derived from the resistant VL51 were treated with 5 µM of ibrutinib (equivalent to the IC50 of the parental and an inactive dose in these cells). In parallel, the same number of cells were exposed to DMSO to highlight molecules relevant to cell survival independent of the drug response.

To achieve 100x coverage of the library, at least 5 µg of gDNA per sample was amplified using appropriate primers flanking the construct region containing the sgRNA. PCR products were purified and then sequenced by amplicon-based next-generation sequencing, 75 bp single-end, using NextSeq500 sequencer. The sgRNAs were mapped, and their distribution was analyzed in all the experimental groups and in the baseline controls. MAGeCKFlute software [84] was applied to perform a quality check of the CRISPRi screen.

Lentivirus and cell line production for competition assay

pgRNAs targeting TSS of elncRNAs selected for further validation were cloned in the pDECKO mCherry vector (RRID:Addgene_78534) as previously described [82]. All the pgRNA guides for CRISPRi validations are reported in Table S12. Lentiviral supernatant for each pgRNA construct was produced as described above. After 72h infection with pgRNA-pDECKO lentiviral supernatants, parental or ibrutinib-resistant VL51 or Jeko1 cells were selected with puromycin (1.5 mg/mL) for two days.

Cells stably expressing pgRNAs and mCherry were then infected with dCas9-Zim3-GFP (RRID: Addgene_188778). After 72 h, cells were analyzed by FACS to determine the percentage of mCherry-GFP double-positive cells at T0. Then, the cells were split into equal amounts and exposed to DMSO or ibrutinib, and tested every three days to measure the survival of double-positive cells compared to cells expressing only pgRNAs. After 14 days double-positive cells were sorted to obtain pure polyclonal populations, to test for ibrutinib sensitivity and for transcriptomic analysis.

The polyclonal populations were also seeded at limiting dilutions to obtain monoclonal cell lines, which were screened for representativeness of the polyclonal cell lines and used for additional validation experiments.

A-to-I editing analysis from tot RNAseq

To analyze A-to-I RNA-editing events, we utilized the RED-ML algorithm [49], a method specifically designed for RNA-editing detection. The analysis began with the aligned sequencing data in BAM format, ensuring that high-quality mapped reads were used as input. RED-ML systematically identified putative RNA-editing sites by distinguishing true A-to-I editing events from sequencing artifacts and single nucleotide polymorphisms (SNPs), leveraging its model for increased accuracy and specificity. Following the identification of RNA-editing events, we annotated the edited sites with gene annotations to associate each editing event with its corresponding gene. Additionally, we classified the events based on their genomic context, determining whether they occurred within gene bodies (3'UTR, 5'UTR, exonic

or intronic regions) or in promoter/downstream/intergenic regions. To further assess the functional
implications of these RNA-editing events, particularly in protein-coding genes, we utilized ANNOVAR to
evaluate their potential impact on coding sequences. Also, editing sites were annotated based on
genomic location, and if they belonged to an Alu sequence or not. Editing sites were considered
sequence-specific if the same nucleotide was edited in at least three samples. Differential RNA editing
analysis was performed using the voom/limma R package [85]. Sequence-specific edited sites were
considered for further analyses. Differentially edited sites were defined as those with an empirical Bayes
corrected (Benjamini-Hockberg procedure) p-value <0.05.

Exon Intron Splicing Analysis

837 Post-transcriptional modulation of transcripts was estimated as previously described [86]

RNA fractionation

839 RNA fractionation was performed as previously described [18]. Briefly, 5 million cells were lysed with RSB-
840 100 buffer containing digitonin to permeabilize cells and recover cytoplasmic RNA. Nuclei were lysed by
841 RSB-100 buffer after the addition of 0.5% Triton-X to collect nuclear RNA by ultracentrifugation. The
842 insoluble fraction was resuspended in nuclear lysis buffer and sonicated to extract chromatin-associated
843 RNA.

DRS

To identify full-length spliced elncRNAs and alternative protein-coding transcripts at a genome-wide
level, all nuclear transcripts, polyadenylated or not, were enriched. 3 µg of RNA extracted from two
monoclonal ADARreg_kd and two negative control KLK12_kd ibrutinib-resistant cell lines were used for
ribodepletion by NEBNext rRNA Depletion Kit. Then, synthetic poly-A tails were added to non-
polyadenylated transcripts by NEB E. coli Poly(A) Polymerase reagent, for library preparation
compatibility. For direct sequencing of native RNA transcripts and modifications, libraries were prepared
with the PCR-free Direct RNA Sequencing (DRS) Kit (SQK-RNA004, Oxford Nanopore) starting from 300
ng of RNA, purified after the previous enzymatic steps and run by Promethion 24 (Oxford Nanopore),
two biological replicates per condition. Around 30 million reads per sample were collected, running a
single RNA sample per flow cell. Similarly, cytosolic RNA was used for library preparation but starting
from 1 µg of unprocessed cytosolic fractions.

Oxford Nanopore Technologies (ONT) long-read sequencing data were processed using Dorado
(https://github.com/nanoporetech/dorado). In particular, Dorado was used for basecalling and for
generating POD5 files directly from raw signals, followed by read alignment to the human reference
genome (GRCh38/hg38) using the integrated alignment module, which internally utilizes minimap2
(https://github.com/lh3/minimap2). Dorado's model-based detection framework was also applied to call
RNA modifications, specifically N6-methyladenosine (m^6A) and adenosine-to-inosine (A-to-I) edits using
the rna004_130bps_hac@v5.2.0 model. To assign modification events to transcript isoforms, we
developed an in-house script that links modification calls to isoform-resolved alignments. Gene- and
isoform-level quantification, as well as de novo transcript discovery, were performed using Bambu [87],
which provides an integrated approach for long-read–based transcript reconstruction and abundance
estimation.

Quantitative Reverse Transcriptase Polymerase Chain Reaction (qRT-PCR)

Strand-specific quantitative RT-PCR (qRT-PCR) was performed using SuperScript III Platinum SYBR Green
One-Step qRT-PCR Kit (Invitrogen). Data were analyzed according to the manufacturer's instructions and
the comparative CT method (ΔΔCT method) then normalized to GAPDH or β-actin as reference genes.
Statistical significance was determined using a two-tailed t-test with a threshold of p< 0.05. Primer
sequences are reported in Supplementary Table S12.

Immunoblotting

Protein extraction, separation, and immunoblotting were performed as previously described.
Nuclear/cytosol fractionation was performed by Nuclear/Cytosol fractionation kit (BioVision, Cat# K266-
100) according manufacturer instructions. The following antibodies were used: anti-ADAR1 (D7E2M)
(Cell Signaling Technology Cat# 14175, RRID:AB_2722520), Anti ADARB1 (Sigma-Aldrich Cat#
SAB1405426, RRID:AB_10740297), anti-IRF3 (D-3) (Santa Cruz Biotechnology Cat# sc-376455,
RRID:AB_11151578) anti-NF-κB p65 (C22B4) (Cell Signaling Technology Cat# 4764, RRID:AB_823578),
anti-phospho-NFkB p65 (Ser536) (Cell Signaling Technology Cat# 3033, RRID:AB_331284), anti-vimentin,
anti-GAPDH (FF26A) from eBioscience, secondary mouse (NA931V) and rabbit (NA934V) antibodies from
GE healthcare. The data were analyzed using Fusion Solo software.

Immunoprecipitation

Nuclei from 10 million cells were isolated using RSB-100 buffer containing digitonin and a phosphatase
and protease inhibitor cocktail. They were then washed and lysed in 1 mL of 50 mM Tris (pH 7.5), 150
mM NaCl, 1% Triton X-100, 1 mM EDTA, and 1 mM EGTA. The insoluble fraction was removed by
centrifugation, 25 μL of the supernatant saved as input, and the rest split in two for the overnight
incubation with 5 μg antibody RPS3 (Proteintech-66046-1-IgG) or 5 ug μg of mouse IgG2a isotype control
(Proteintech 66360-2-IgG-150UL). The day after protein complexes were recovered by magnetic protein
A beads (20 μL, 2h 4°C in rotation), then washed three times with lysis buffer and then protein
complexes were eluted in 20 μL of Laemmli buffer, boiled 10' at 95°C and separated as previously
described. For the blotting anti-phospho-NFkB p65 (Ser536) (Cell Signaling Technology Cat# 3033,
RRID:AB_331284) and RPS3 (Proteintech-66046-1-IgG) were used and detected by Clean-Blot IP
Detection Reagent (HRP) (Thermo Scientific™, cat #21230) to avoid the cross reactivity of the heavy and
light chains of the bait antibody.

Immunofluorescence and confocal microscopy

Ibrutinib resistant VL51 cells (100,000 cells/well) were seeded on poly-L-ornithine coated 18-well IBIDI
chambers (81816, IBIDI GmbH), then fixed for 20 min with Paraformaldehyde (PFA) 4% at room
temperature (RT). Cells were permeabilized with PBS + 0.1% Triton X-100 for 10 min at RT. To block
unspecific staining, samples were blocked for 1 h with 5% Bovine Serum Albumin (BSA) at RT before
staining. Antibodies were diluted in 5% BSA. Samples were incubated overnight at 4°C with primary
antibody anti IRF3 (D-3) (Santa Cruz Biotechnology Cat# sc-376455, RRID:AB_11151578), anti-ADARB1
(Sigma-Aldrich Cat# SAB1405426, RRID:AB_10740297), anti-ADAR1 (Santa Cruz Biotechnology Cat# sc-
73408, RRID:AB_2222767), or anti-NF-κB p65 (C22B4) (Cell Signaling Technology Cat# 4764,
RRID:AB_823578). Secondary goat antibody anti-mouse IgG labelled with Alexa647 (A-21235, Thermo
Fisher Scientific Cat# A-21235, RRID:AB_2535804) 1 h at RT in the dark. Slides were counterstained after
3 washes of PBS with 0.3 μg/mL 4,6-diamidino-2-phenylindole (Sigma-Aldrich). Images were acquired
on a Leica SP5 with an objective with ×63 magnification. Protein quantification and nuclear/ cytosol ratio

were evaluated by ImageJ (RRID:SCR_003070) software. At least three images per condition, including
approximately 30-50 cells each, were used for each quantification.

Co-culture with NK cells

PBMCs were collected from healthy donors through density gradient separation (Ficoll Paque Plus,
Merck). NK isolation was performed using the Easy-Sep Human NK Cell Isolation kit (STEMCELL
Technologies) following manufactural instructions. Freshly isolated NK cells were cultured with RPMI
1640 medium supplemented with IL-2 (2.5ng/ml; PeproTech) and IL-15 (10ng/ml; PeproTech).

Single-cell RNAseq data

The single-cell sequencing dataset (GSE111014) was obtained from GEO and is associated with the
referenced publication [51]. The re-analysis of the dataset files (GSE111014_barcodes.tsv.gz,
GSE111014_genes.tsv.gz, and GSE111014_matrix.mtx.gz) was conducted from scratch using the Seurat
package in R. The raw data were pre-processed to generate a Seurat object, followed by quality control,
normalization, and feature selection steps to identify variable genes. Dimensionality reduction was
performed using principal component analysis (PCA) and visualized through uniform manifold
approximation and projection (UMAP). For cell-type annotation, the SingleR algorithm was applied,
leveraging reference datasets to predict cell identities. The output of SingleR was then subjected to a
manual revision to refine and ensure the accuracy of the annotations, combining algorithmic insights
with domain expertise. B cells were subdivided in ADARreg ^{high}-like or ADARreg ^{low}-like based on the
signature score calculated by R package AUCell using IBRr VL51 ADARreg knockdown differentially
expressed genes. NK cell subsets were defined based on cytotoxicity, inflammatory and stress-related
gene sets as previously described[52]. Cell-cell interactions at pretreatment time points in patient CLL8 and
CLL6 were analyzed by CellChat.

Data mining

Functional analysis was performed using GSEA[88] (Gene Set Enrichment Analysis) on genes ranked by t-
statistic as calculated by Limma test from differential gene expression analysis. Gene sets were
considered significantly enriched if FDR<0.05.

Gene ontology analysis was performed using the g-Profiler[89] webtool. The p-value for pathway
enrichment was computed using a Fisher's exact test multiple-test correction was applied.

Functional impact of differential isoform expression was performed by tappAS[32] software using a
functional iso-transcriptomic approach to retrieve functional annotation features at both RNA and
protein levels.

Semantic clustering analysis was performed on long lists of gene ontology terms using Revigo[90].

Statistical methods were implemented in R and executed within the Rscript environment.

Supplemental information

Figure S1 (related to Figure 2): Functional screening of elncRNAs linked to BCR blockade

Figure S2 (related to Figure 2): lncPCED1b and ADARreg elncRNAs genomic loci

Figure S3 (related to Figure 3): lncPCED1b and ADARreg knockdown biological effects

Figure S4 (related to Figure 4): A-to-I RNA editing evaluated by total RNAseq

Figure S5 (related to Figure 5): Transcriptional and post-transcriptional impact of lncPCED1b and
ADARreg silencing evaluated by total RNAseq.

Figure S6 (related to Figure 6): Transcriptional and post-transcriptional impact of ADARreg silencing
evaluated by DRS on subcellular compartments.

Figure S7 (related to Figure 7): Modulation of key RNA isoforms of the immune-suppressive
transcriptional program of lymphomas by ADARreg.

Figure S8 (related to Figure 7): Crosstalk between NK cell and tumor correlated with ADARreg signature
in CLL patients.

Figure S9 (related to Figure 7): proposed model

Table S1 (separate file, related to Fig.1): VL51_Ibru vs parental VL51 specific enhancers and DEGs

Ibrutinib-resistant versus parental VL51 specific enhancers detected by ChIPseq and annotated
with differentially expressed genes, resulting from the comparison of gene expression profiles of
VL51-Ibru versus parental VL51 and falling within a 40kb window from the peak. LogFC represents
modulation for each gene in ibrutinib-resistant vs parental.

Table S2 (separate file, related to Fig.2): Composition of elncRNA SMZL CRISPR-i library

Composition of elncRNA SMZL CRISPR-i library**.** Sequences of both sgRNAs of the pair binding
each target with relative chromosomal position are listed.

Table S3 (separate file, related to Fig.2): CRISPRi screen

CRISPRi screen enrichment scores (beta) of each sample respect the time 0. Targets are
considered statistically significant enriched for fdr<0.2 (green)

Table S4: **(separate file, related to Fig.4): A-I-RNA editing**

Limma test performed on A-to-I RNA editing for each edited site detected by RED-ML. LogFC
represents modulation for each editing site in ibrutinib-resistant vs parental cells (RESacq) or in
parental (PAR) or ibrutinib-resistant (RES) cells knockdown for elncRNA of interest vs the negative
control.

Table S5 (separate file, related to Fig.5): DEG ADARreg_lncPCED1b_vs_KLKL12kd

Limma test performed on gene expression profiles of ibrutinib-resistant vs parental cells
(IBRr_KLK12kd vs PAR_KLK12kd) or in ibrutinib-resistant cells knockdown for ADARreg or
lncPCED1b vs the negative control. LogFC represents modulation for each significantly
differentially expressed gene (adj.P<0.05) in the indicated comparison
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Table S6 (separate file, related to Fig.5): GSEA

Genesets significantly enriched in at least one of the following comparisons: ibrutinib-resistant
versus parental cells, ibrutinib-resistant ADARreg or lncPCED1b knockdown versus negative
control. Enrichment score (NES) and fdr calculated by GSEA software are indicated for each
geneset and each comparison.

Table S7 (separate file, related to Fig.5): Post-transcriptionally modulated transcripts_EISA

D exons and D introns measured by EISA in the comparison of ibrutinib-resistant cells with
ADARreg or lncPCED1b knockdown versus negative control. Transcripts are considered post-
transcriptionally modulated if Dex/Dint in ADARreg or lncPCED1b vs KLK12_kd   have p<0.05 in
Limma test.

Table S8 (separate file, related to Fig.S6): editing of alternative transcripts

List of genes with transcriptional isoforms modulated in opposite direction (opposite LogFC and adj
p-value <0.05 in Limma test performed on quantification per isoforms of total RNAseq in ibrutinib-
resistant ADARreg or lncPCED1b knockdown versus control cells). Additional information about
significantly differential intronic or 3' UTR editing events is provided.

Table S9 (separate file, related to Fig.6): DIU DEA summary

TappAS summary of differential isoform usage (DIU) and differential expression analysis (DEA) in
nuclear transcripts quantified by full-length reads in DRS.

Table S10 (related to STAR Methods): oligos for CRISPRi sequencing and validation experiments

Sequences of pgRNAs cloned in lentiviral vector for stable silencing of elncRNAs, antisense
oligonucleotides used for transient silencing, primers to prepare CRISPRi library amplicons for
sequencing, and primers to detect transcript expression by qRT.

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References

1.	Rossi D, Bertoni F, Zucca E. Marginal-Zone Lymphomas. *N Engl J Med*. 2022;386(6):568-581.

2.	Laurent C, Bertoni F. The Biology of MZL subtypes: Challenge and Relevance of Classification.
Blood. 2025.

3.	Petro JB, Rahman SM, Ballard DW, Khan WN. Bruton's tyrosine kinase is required for activation
of IkappaB kinase and nuclear factor kappaB in response to B cell receptor engagement. *J Exp Med*.
2000;191(10):1745-1754.

4.	Stanchina MD, Montoya S, Danilov AV, et al. Navigating the changing landscape of BTK-targeted
therapies for B cell lymphomas and chronic lymphocytic leukaemia. *Nat Rev Clin Oncol*.
2024;21(12):867-887.

5.	Cool A, Nong T, Montoya S, Taylor J. BTK inhibitors: past, present, and future. *Trends Pharmacol*
Sci. 2024;45(8):691-707.

6.	Mouhssine S, Maher N, Matti BF, Alwan AF, Gaidano G. Targeting BTK in B Cell Malignancies:
From Mode of Action to Resistance Mechanisms. *Int J Mol Sci*. 2024;25(6).

7.	Nakhoda S, Vistarop A, Wang YL. Resistance to Bruton tyrosine kinase inhibition in chronic
lymphocytic leukaemia and non-Hodgkin lymphoma. *Br J Haematol*. 2023;200(2):137-149.

8.	Shaffer AL, 3rd, Phelan JD, Wang JQ, et al. Overcoming Acquired Epigenetic Resistance to BTK
Inhibitors. *Blood Cancer Discov*. 2021;2(6):630-647.

9.	Zhao X, Wang MY, Jiang H, et al. Transcriptional programming drives Ibrutinib-resistance
evolution in mantle cell lymphoma. *Cell Rep*. 2021;34(11):108870.

10.	Arribas AJ, Napoli S, Cascione L, et al. ERBB4-Mediated Signaling Is a Mediator of Resistance to
PI3K and BTK Inhibitors in B-cell Lymphoid Neoplasms. *Mol Cancer Ther*. 2024;23(3):368-380.

11.	Arribas AJ, Napoli S, Cascione L, et al. Resistance to PI3Kdelta inhibitors in marginal zone
lymphoma can be reverted by targeting the IL-6/PDGFRA axis. *Haematologica*. 2022;107(11):2685-2697.

12.	Mattick JS, Amaral PP, Carninci P, et al. Long non-coding RNAs: definitions, functions, challenges
and recommendations. *Nat Rev Mol Cell Biol*. 2023;24(6):430-447.

13.	Saha D, Animireddy S, Lee J, et al. Enhancer switching in cell lineage priming is linked to eRNA,
Brg1's AT-hook, and SWI/SNF recruitment. *Mol Cell*. 2024;84(10):1855-1869 e1855.

14.	Andersson R, Gebhard C, Miguel-Escalada I, et al. An atlas of active enhancers across human cell
types and tissues. *Nature*. 2014;507(7493):455-461.

15.	Lai F, Orom UA, Cesaroni M, et al. Activating RNAs associate with Mediator to enhance
chromatin architecture and transcription. *Nature*. 2013;494(7438):497-501.

16.	Song P, Han R, Yang F. Super enhancer lncRNAs: a novel hallmark in cancer. *Cell Commun Signal*.
2024;22(1):207.

17.	Agirre X, Meydan C, Jiang Y, et al. Long non-coding RNAs discriminate the stages and gene
regulatory states of human humoral immune response. *Nat Commun*. 2019;10(1):821.

18.	Napoli S, Cascione L, Rinaldi A, et al. Characterization of GECPAR, a noncoding RNA that
regulates the transcriptional program of diffuse large B-cell lymphoma. *Haematologica*.
2022;107(5):1131-1143.

19.	Coan M, Haefliger S, Ounzain S, Johnson R. Targeting and engineering long non-coding RNAs for
cancer therapy. *Nat Rev Genet*. 2024;25(8):578-595.

20.	Giacopuzzi E, Gennarelli M, Sacco C, et al. Genome-wide analysis of consistently RNA edited
sites in human blood reveals interactions with mRNA processing genes and suggests correlations with
cell types and biological variables. *BMC Genomics*. 2018;19(1):963.

21.	Arribas A, Napoli S, Cascione L, et al. Il16 Production Is a Mechanism of Resistance to Btk
Inhibitors and to R-Chop. *Hematological Oncology*. 2023;41(S2):188-189.

22.	Inokuchi K, Abo J, Takahashi H, et al. Establishment and characterization of a villous lymphoma
cell line from splenic B-cell lymphoma. *Leuk Res*. 1995;19(11):817-822.

23.   Jha S, Pan-Yun Ting J. Holding the inflammatory system in check: NLRs keep it cool. *F1000Prime*
Rep. 2015;7:15.
24.   Chen H, Liang H. A High-Resolution Map of Human Enhancer RNA Loci Characterizes Super-
enhancer Activities in Cancer. *Cancer Cell*. 2020;38(5):701-715 e705.
25.   Whyte WA, Orlando DA, Hnisz D, et al. Master transcription factors and mediator establish
super-enhancers at key cell identity genes. *Cell*. 2013;153(2):307-319.
26.   Consortium F, the RP, Clst, et al. A promoter-level mammalian expression atlas. *Nature*.
2014;507(7493):462-470.
27.   Alerasool N, Segal D, Lee H, Taipale M. An efficient KRAB domain for CRISPRi applications in
human cells. *Nat Methods*. 2020;17(11):1093-1096.
28.   Xiong H, Liu D, Li Q, et al. RED-ML: a novel, effective RNA editing detection method based on
machine learning. *Gigascience*. 2017;6(5):1-8.
29.   Sinigaglia K, Wiatrek D, Khan A, et al. ADAR RNA editing in innate immune response phasing, in
circadian clocks and in sleep. *Biochim Biophys Acta Gene Regul Mech*. 2019;1862(3):356-369.
30.   Li C, Ma L, Chen X. Interferon regulatory factor 3-CL, an isoform of IRF3, antagonizes activity of
IRF3. *Cell Mol Immunol*. 2011;8(1):67-74.
31.   Diensthuber G, Novoa EM. Charting the epitranscriptomic landscape across RNA biotypes using
native RNA nanopore sequencing. *Mol Cell*. 2025;85(2):276-289.
32.   de la Fuente L, Arzalluz-Luque A, Tardaguila M, et al. tappAS: a comprehensive computational
framework for the analysis of the functional impact of differential splicing. *Genome Biol*. 2020;21(1):119.
33.   Kapadia B, Roychowdhury A, Kayastha F, et al. m6A eraser ALKBH5/treRNA1/DDX46 axis
regulates BCR expression. *Neoplasia*. 2025;62:101144.
34.   Dong F, Qin X, Wang B, et al. ALKBH5 Facilitates Hypoxia-Induced Paraspeckle Assembly and IL8
Secretion to Generate an Immunosuppressive Tumor Microenvironment. *Cancer Res*. 2021;81(23):5876-
5888.
35.   Xiao W, Adhikari S, Dahal U, et al. Nuclear m(6)A Reader YTHDC1 Regulates mRNA Splicing. *Mol*
Cell. 2016;61(4):507-519.
36.   Wang ZW, Pan JJ, Hu JF, et al. SRSF3-mediated regulation of N6-methyladenosine modification-
related lncRNA ANRIL splicing promotes resistance of pancreatic cancer to gemcitabine. *Cell Rep*.
2022;39(6):110813.
37.   Ma L, Xue X, Zhang X, et al. The essential roles of m(6)A RNA modification to stimulate ENO1-
dependent glycolysis and tumorigenesis in lung adenocarcinoma. *J Exp Clin Cancer Res*. 2022;41(1):36.
38.   Rong S, Dai B, Yang C, et al. HNRNPC modulates PKM alternative splicing via m6A methylation,
upregulating PKM2 expression to promote aerobic glycolysis in papillary thyroid carcinoma and drive
malignant progression. *J Transl Med*. 2024;22(1):914.
39.   Tan L, Cheng W, Liu F, et al. Positive natural selection of N6-methyladenosine on the RNAs of
processed pseudogenes. *Genome Biol*. 2021;22(1):180.
40.   Yang X, Triboulet R, Liu Q, Sendinc E, Gregory RI. Exon junction complex shapes the m(6)A
epitranscriptome. *Nat Commun*. 2022;13(1):7904.
41.   Luo Z, Ma Q, Sun S, et al. Exon-intron boundary inhibits m(6)A deposition, enabling m(6)A
distribution hallmark, longer mRNA half-life and flexible protein coding. *Nat Commun*. 2023;14(1):4172.
42.   Camponeschi A, Klasener K, Sundell T, et al. Human CD38 regulates B cell antigen receptor
dynamic organization in normal and malignant B cells. *J Exp Med*. 2022;219(9).
43.   Liu PC, Huang SY, Lin KI, Hsieh SL, Leu CM. Suppression of NF-kappaB and downstream XBP1 by
DcR3 contributes to a decrease in antibody secretion. *J Immunol*. 2025;214(1):72-84.
44.   Ozga AJ, Chow MT, Luster AD. Chemokines and the immune response to cancer. *Immunity*.
2021;54(5):859-874.

45. Xavier S, Nguyen V, Khairnar V, et al. CEACAM1 as a mediator of B-cell receptor signaling in
mantle cell lymphoma. *Nat Commun*. 2025;16(1):4967.
46. Helfrich I, Singer BB. Size Matters: The Functional Role of the CEACAM1 Isoform Signature and
Its Impact for NK Cell-Mediated Killing in Melanoma. *Cancers (Basel)*. 2019;11(3).
47. Braud VM, Meghraoui-Kheddar A, Elaldi R, Petti L, Germain C, Anjuere F. LLT1-CD161 Interaction
in Cancer: Promises and Challenges. *Front Immunol*. 2022;13:847576.
48. Germain C, Bihl F, Zahn S, et al. Characterization of alternatively spliced transcript variants of
CLEC2D gene. *J Biol Chem*. 2010;285(46):36207-36215.
49. Wan F, Anderson DE, Barnitz RA, et al. Ribosomal protein S3: a KH domain subunit in NF-kappaB
complexes that mediates selective gene regulation. *Cell*. 2007;131(5):927-939.
50. Jang CY, Lee JY, Kim J. RpS3, a DNA repair endonuclease and ribosomal protein, is involved in
apoptosis. *FEBS Lett*. 2004;560(1-3):81-85.
51. Rendeiro AF, Krausgruber T, Fortelny N, et al. Chromatin mapping and single-cell immune
profiling define the temporal dynamics of ibrutinib response in CLL. *Nat Commun*. 2020;11(1):577.
52. Tang F, Li J, Qi L, et al. A pan-cancer single-cell panorama of human natural killer cells. *Cell*.
2023;186(19):4235-4251 e4220.
53. Li Q, Yang Z, He X, Yang X. Comprehensive analysis of PILRAlpha's association with the prognosis,
tumor immune infiltration, and immunotherapy in pan-cancer. *Sci Rep*. 2023;13(1):14334.
54. Lanng KRB, Lauridsen EL, Jakobsen MR. The balance of STING signaling orchestrates immunity in
cancer. *Nat Immunol*. 2024;25(7):1144-1157.
55. Tan JY, Marques AC. The activity of human enhancers is modulated by the splicing of their
associated lncRNAs. *PLoS Comput Biol*. 2022;18(1):e1009722.
56. Levanon EY, Cohen-Fultheim R, Eisenberg E. In search of critical dsRNA targets of ADAR1. *Trends
Genet*. 2024;40(3):250-259.
57. Li Q, Gloudemans MJ, Geisinger JM, et al. RNA editing underlies genetic risk of common
inflammatory diseases. *Nature*. 2022;608(7923):569-577.
58. Wang IX, So E, Devlin JL, Zhao Y, Wu M, Cheung VG. ADAR regulates RNA editing, transcript
stability, and gene expression. *Cell Rep*. 2013;5(3):849-860.
59. Shevchenko G, Morris KV. All I's on the RADAR: role of ADAR in gene regulation. *FEBS Lett*.
2018;592(17):2860-2873.
60. Tang SJ, Shen H, An O, et al. Cis- and trans-regulations of pre-mRNA splicing by RNA editing
enzymes influence cancer development. *Nat Commun*. 2020;11(1):799.
61. Hajji K, Sedmik J, Cherian A, Amoruso D, Keegan LP, O'Connell MA. ADAR2 enzymes: efficient
site-specific RNA editors with gene therapy aspirations. *RNA*. 2022;28(10):1281-1297.
62. Li Z, Tian Y, Tian N, et al. Aberrant alternative splicing pattern of ADAR2 downregulates
adenosine-to-inosine editing in glioma. *Oncol Rep*. 2015;33(6):2845-2852.
63. Sciarrillo R, Wojtuszkiewicz A, Assaraf YG, et al. The role of alternative splicing in cancer: From
oncogenesis to drug resistance. *Drug Resist Updat*. 2020;53:100728.
64. Sun K, Li J, Chen C, et al. Advances in nanopore direct RNA sequencing and its impact on
biological research. *Biotechnol Adv*. 2025;85:108710.
65. Xie YY, Zhong ZD, Chen HX, et al. Single-molecule direct RNA sequencing reveals the shaping of
epitranscriptome across multiple species. *Nat Commun*. 2025;16(1):5119.
66. Roundtree IA, Luo GZ, Zhang Z, et al. YTHDC1 mediates nuclear export of N(6)-methyladenosine
methylated mRNAs. *Elife*. 2017;6.
67. Kelemen K. The Role of Nucleophosmin 1 (NPM1) Mutation in the Diagnosis and Management of
Myeloid Neoplasms. *Life (Basel)*. 2022;12(1).
68. Lin J, Kato M, Nagata K, Okuwaki M. Efficient DNA binding of NF-kappaB requires the chaperone-
like function of NPM1. *Nucleic Acids Res*. 2017;45(7):3707-3723.

69.   Wang X, Chai Y, Quan Y, et al. NPM1 inhibits tumoral antigen presentation to promote immune
evasion and tumor progression. *J Hematol Oncol*. 2024;17(1):97.
70.   Nakagawa S, Naganuma T, Shioi G, Hirose T. Paraspeckles are subpopulation-specific nuclear
bodies that are not essential in mice. *J Cell Biol*. 2011;193(1):31-39.
71.   Clemson CM, Hutchinson JN, Sara SA, et al. An architectural role for a nuclear noncoding RNA:
NEAT1 RNA is essential for the structure of paraspeckles. *Mol Cell*. 2009;33(6):717-726.
72.   Imamura K, Imamachi N, Akizuki G, et al. Long noncoding RNA NEAT1-dependent SFPQ
relocation from promoter region to paraspeckle mediates IL8 expression upon immune stimuli. *Mol Cell*.
2014;53(3):393-406.
73.   Loucas G, Locker N, Parker R. Nucleic acid-protein condensates in innate immunity. *Mol Cell*.
2025;85(20):3823-3839.
74.   Wier EM, Neighoff J, Sun X, Fu K, Wan F. Identification of an N-terminal truncation of the NF-
kappaB p65 subunit that specifically modulates ribosomal protein S3-dependent NF-kappaB gene
expression. *J Biol Chem*. 2012;287(51):43019-43029.
75.   Gao X, Hardwidge PR. Ribosomal protein s3: a multifunctional target of attaching/effacing
bacterial pathogens. *Front Microbiol*. 2011;2:137.
76.   Bonfiglio F, Bruscaggin A, Guidetti F, et al. Genetic and phenotypic attributes of splenic marginal
zone lymphoma. *Blood*. 2022;139(5):732-747.
77.   Kotlov N, Bagaev A, Revuelta MV, et al. Clinical and Biological Subtypes of B-cell Lymphoma
Revealed by Microenvironmental Signatures. *Cancer Discov*. 2021;11(6):1468-1489.
78.   Guidetti F, Arribas AJ, Sartori G, et al. Targeting IRAK4 with Emavusertib in Lymphoma Models
with Secondary Resistance to PI3K and BTK Inhibitors. *J Clin Med*. 2023;12(2).
79.   Bolger AM, Lohse M, Usadel B. Trimmomatic: a flexible trimmer for Illumina sequence data.
Bioinformatics. 2014;30(15):2114-2120.
80.   Dobin A, Davis CA, Schlesinger F, et al. STAR: ultrafast universal RNA-seq aligner. *Bioinformatics*.
2013;29(1):15-21.
81.   Esposito R, Polidori T, Meise DF, et al. Multi-hallmark long noncoding RNA maps reveal non-
small cell lung cancer vulnerabilities. *Cell Genom*. 2022;2(9):100171.
82.   Pulido-Quetglas C, Aparicio-Prat E, Arnan C, et al. Scalable Design of Paired CRISPR Guide RNAs
for Genomic Deletion. *PLoS Comput Biol*. 2017;13(3):e1005341.
83.   McDonald ER, 3rd, de Weck A, Schlabach MR, et al. Project DRIVE: A Compendium of Cancer
Dependencies and Synthetic Lethal Relationships Uncovered by Large-Scale, Deep RNAi Screening. *Cell*.
2017;170(3):577-592 e510.
84.   Wang B, Wang M, Zhang W, et al. Integrative analysis of pooled CRISPR genetic screens using
MAGeCKFlute. *Nat Protoc*. 2019;14(3):756-780.
85.   Law CW, Chen Y, Shi W, Smyth GK. voom: Precision weights unlock linear model analysis tools
for RNA-seq read counts. *Genome Biol*. 2014;15(2):R29.
86.   Munz N, Cascione L, Parmigiani L, et al. Exon-Intron Differential Analysis Reveals the Role of
Competing Endogenous RNAs in Post-Transcriptional Regulation of Translation. *Noncoding RNA*.
2021;7(2).
87.   Chen Y, Sim A, Wan YK, et al. Context-aware transcript quantification from long-read RNA-seq
data with Bambu. *Nat Methods*. 2023;20(8):1187-1195.
88.   Subramanian A, Tamayo P, Mootha VK, et al. Gene set enrichment analysis: a knowledge-based
approach for interpreting genome-wide expression profiles. *Proc Natl Acad Sci U S A*.
2005;102(43):15545-15550.
89.   Reimand J, Arak T, Adler P, et al. g:Profiler-a web server for functional interpretation of gene
lists (2016 update). *Nucleic Acids Res*. 2016;44(W1):W83-89.

90.   Supek F, Bosnjak M, Skunca N, Smuc T. REVIGO summarizes and visualizes long lists of gene
ontology terms. *PLoS One*. 2011;6(7):e21800.

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