BRD2 upregulation as a pan-cancer adaptive resistance mechanism to BET inhibition

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Abstract

Bromodomain and extraterminal motif (BET) inhibitors, such as JQ1, are promising cancer

therapeutics that target epigenetic regulators, particularly BRD4. However, resistance to BET

inhibitors (BETi) limits their clinical utility, necessitating a better understanding of adaptive

mechanisms. We identified BRD2 upregulation as a conserved response to BET inhibition across

multiple cancer types and hypothesized that BRD2 compensates for BRD4 loss, sustaining

essential transcriptional programs upon treatment. Consistent with this, BRD2 knockdown

sensitized cancer cells to BETi *in vitro*, and combining BRD2 depletion and JQ1 treatment

significantly impaired tumor growth *in vivo*. At the chromatin level, BRD2 and BRD4 ChIP-seq

analysis of pancreatic cancer cells showed consistent BRD4 loss from chromatin after JQ1

treatment, while BRD2 displacement differed by sensitivity. Resistant cells maintained higher

BRD2 occupancy than sensitive cells, suggesting a link between BRD2 retention and drug

response. Mechanistically, NFYA mediates BRD2 upregulation as NFYA depletion attenuated

BRD2 upregulation upon BETi treatment. Collectively, our findings establish BRD2 as a critical

mediator of pan-cancer adaptive resistance to BETi and identify NFYA as a novel transcriptional

regulator of this process. Co-targeting BRD2 or its regulatory network offers a rational strategy to

enhance the durability and efficacy of BET-based therapies.

Keywords

BET inhibitor, resistance, pan-cancer, BRD2, BRD4

Introduction

Cancer progression can be driven by epigenetic reprogramming, which has recently been recognized as a fundamental hallmark of cancer (1, 2). Epigenetic alterations enable cancer cells to activate oncogenic programs, silence tumor suppressors and adapt to microenvironmental stress, thereby promoting cancer progression and metastasis (3). Importantly, while conferring plasticity and survival, these changes render cancer cells dependent on aberrant epigenetic regulation to sustain the malignant phenotype—vulnerabilities that can be therapeutically exploited. Central to this process is the molecular machinery that regulates epigenetic states, coordinated by three major classes of proteins: writers, readers, and erasers. Writers deposit modifications such as DNA methylation or histone acetylation, erasers remove them, and readers interpret these marks to regulate downstream transcriptional activity. Together, these mechanisms dynamically shape transcriptional programs, often hijacked during carcinogenesis.

Among these regulators, the BET (bromodomain and extra-terminal domain) family has drawn significant attention as a group of critical epigenetic readers. This family comprises four members—BRD2, BRD3, BRD4, and the testis-specific BRDT—all sharing a key structural feature: two tandem N-terminal bromodomains that recognize and bind acetylated lysine residues on histones to anchor transcriptional machinery at active chromatin (4, 5). Of these, BRD4 is the most extensively studied and plays a critical role at enhancer and super-enhancers (SE), large regulatory clusters that drive the expression of key oncogenes, such as *MYC* (6). In addition to tethering to acetylated chromatin, BRD4 also recruits and scaffolds the Mediator complex and the P-TEFb kinase to release paused RNA polymerase II, thereby sustaining high-level transcription of target genes (7). Collectively, the actions of BRD4 support self-reinforcing transcriptional programs driven by inherently unstable oncogenic *MYC* transcripts (3). Because direct pharmacologic inhibition of MYC has proven difficult, BRD4 represents an attractive therapeutic vulnerability. Small-molecule BET inhibitors (BETi) such as JQ1 competitively bind to the

bromodomains—most potently of BRD4—displacing them from chromatin and dismantling SE-
driven transcription (8). This results in rapid *MYC* downregulation and potent antitumor effects in
preclinical models, with notable success in MYC-addicted hematologic malignancies and an
increasing exploration across diverse cancer types (9-11).

Despite these promising preclinical outcomes, clinical responses to BETi have been modest and
short-lived, with resistance posing as a major barrier for translation (12-14). To date, most
mechanistic insights into BETi resistance have come from models of acquired resistance following
prolonged drug exposure, in which cancer cells undergo extensive transcriptional and epigenetic
rewiring. Such models may not reflect the clinically relevant dynamics of adaptive resistance,
defined by early and reversible changes that enable cancer cells to tolerate BET inhibition. A
deeper understanding of the early adaptive responses to BET inhibition could reveal
vulnerabilities that precede the emergence of acquired resistance and guide the development of
more effective upfront or combination therapies.

In this study, we uncovered upregulation of BRD2—a BRD4 paralog—as an adaptive response
to BET inhibition across multiple cancer types. Functionally, BRD2 knockdown (KD) sensitizes
tumors to the BETi JQ1, enhancing therapeutic efficacy in both *in vitro* and *in vivo* models.
Mechanistically, this upregulation is driven, at least in part, by increased expression of the
transcription factor NFYA. Notably, our results indicate that the extent of the initial BETi response
could help predict the benefit of combining BRD2 KD with JQ1. While all BETi-treated cells benefit
from the combination, the most substantial enhancements are observed in initially resistant
populations. These findings provide a rationale for incorporating BRD2 targeting into combination
strategies to enhance BETi potency and prolong their clinical benefit across diverse cancers.

Materials and methods

Cell lines and cell culture conditions

A diverse panel of mouse and human cancer cell lines was used in this study. All cells were

maintained at 37°C in a humidified incubator with 5% CO_2 under standard conditions appropriate

for each cell type. The base culture media included Dulbecco's Modified Eagle Medium (DMEM;

Corning, 10-013-CV), RPMI-1640 (Gibco, 11875119), Iscove's Modified Dulbecco's Medium

(IMDM; Gibco, 12440053), McCoy's 5A Modified Medium (Gibco, 16600082), Eagle's Minimum

Essential Medium (EMEM; ATCC, 30-2003), and Advanced DMEM/F-12 (Gibco, 12634028).

Unless otherwise specified, culture media were supplemented with 10% fetal bovine serum

(FBS; GenClone, 25-550H) and 1% penicillin-streptomycin (P/S; Gibco, 15140-122). Triple-

negative breast cancer (TNBC) cell lines were additionally supplemented with 10 mM HEPES

(Gibco, 15630080), 1 µg/ml hydrocortisone, and 5 µg/ml insulin (Sigma-Aldrich, I0516-5ML). Cell

line information, including species, tumor type, tissue origin, specific culture conditions, and

corresponding sources and catalog numbers, is listed in the Table S1.

Drug compounds and treatment conditions

All compounds were purchased from the suppliers listed in Table S2 and dissolved in dimethyl

sulfoxide (DMSO; Sigma-Aldrich, 472301-500ML) prior to use as stock solutions. Unless

otherwise specified, treatments were performed at final concentration of 1 µM for 24 hours in

mouse samples and 10 µM for 24 hours in human samples. For control conditions, cells were

treated with DMSO at the same final concentration used to dissolve drug compounds.

Cell viability assay

Mouse cancer cells and A549, U251, and U2OS cells were seeded at 1,500 cells/well, while all

other human cancer cells were seeded at 3,000 cells/well in 50 µl of complete medium into 96-

well plates. Twenty-four hours after cell plating, drugs were serially diluted in complete medium

and 50 µl of the diluted compounds was added to each well. Cells were incubated for an

additional 72 hours before media were carefully removed. To assess cell viability, 100 µl of

diluted AlamarBlue solution prepared by mixing 200 µl of resazurin (Fisher Scientific,

AC418900010) in 50 ml of 1× PBS was added to each well. Plates were incubated at 37°C with

5% CO_2 for 3–4 hours. Absorbance at 570 nm was measured using a plate reader (SpectraMax

iD5 plate reader, Molecular Devices). Relative cell viability was calculated by normalizing the

absorbance of drug-treated wells to DMSO-treated controls. Dose–response curves were

generated using nonlinear regression (curve fitting) in GraphPad Prism version 10.6.0 (796).

Colony formation assay

One thousand cells were seeded per well in 6-well plates. After 1–3 days, cells were treated with

JQ1 (50–100 nM) for 7–10 days, with fresh drug-containing medium replaced every 2–3 days. At

the endpoint, cells were fixed with 100% methanol (15 minutes, room temperature) and stained

with 0.1% (w/v) crystal violet (Spectrum chemical, TCI-C0428-100G) in 20% methanol. Plates

were rinsed under running tap water until excess stain was no longer visible, air-dried

completely, and scanned (HP LaserJet Pro). Colony area and staining intensity were quantified

in ImageJ using the ColonyArea plugin.

shRNA construct generation via pLKO.1 cloning

For the generation of shRNA constructs, the pLKO.1-puro vector (Addgene #8453) was used. A

total of 6 µg of plasmid DNA was digested with 1 µl EcoRI-HF (NEB, R3101T), 1 µl AgeI-HF

(NEB, R3552S), and 5 µl rCutSmart Buffer (NEB, B6004S) in a 50 µl reaction and incubated at

37°C for 3 hours. To prevent vector self-ligation, 1 µl Quick CIP (NEB, M0525S) was added and

incubated for an additional hour at 37°C. The digested vector was purified using the PureLink

PCR purification kit (Invitrogen, K310002). For oligonucleotide preparation, 1 µl of forward and

reverse primers (each at 100 µM) were combined with 1 µl 10× T4 DNA ligase buffer (NEB,

B0202S), 0.5 µl T4 Polynucleotide Kinase (NEB, M0201S), and 6.5 µl nuclease-free water. The

reaction was incubated at 37°C for 30 minutes, then heated to 95 °C for 5 minutes, followed by a

gradual ramp down to 25°C at 5 °C/minute for annealing. The annealed primers were diluted

1:200 before ligation. Ligation reactions were performed by mixing 50 ng of digested pLKO.1

vector, 1 µl diluted annealed oligos, 1 µl 10× T4 ligase buffer (NEB, B0202S), 1 µl T4 DNA ligase

(NEB, M0202S), and water to a final volume of 10 µl. The mixture was incubated at room

temperature for 30 minutes. For transformation, 3 µl of the ligation mix was added to 30 µl XL10-

Gold ultracompetent cells (Agilent, 50-125-094). After 5 minutes on ice, cells were heat-shocked

at 42°C for 90 seconds, returned to ice for another 5 minutes, and plated on LB agar (Fisher

Scientific, BP1427-500) supplemented with 1× ampicillin (RPI, C802G69) using 4.5 mm sterile

rattler plating beads (Zymo Research, S1001). Plates were incubated overnight at 37°C (~16

hours). At least 6 individual colonies were miniprepped using Zyppy Plasmid Miniprep Kit (Zymo

Research, D4019). Plasmid inserts were verified by Sanger sequencing (GENEWIZ) using

primers targeting the U6 promoter. Resulting sequences were aligned to reference in Benchling

to confirm correct oligo insertion. Information about the short hairpin RNA (shRNA) constructs

used in this study, including the target gene, sequence ID, and oligonucleotide sequences, is

summarized in the Table S3.

Lentiviral production and transduction

Lentiviral particles were produced by transfecting HEK293T cells using a standard third-

generation packaging system. Briefly, $5 × 10^6$ HEK293T cells were seeded in 10 cm culture

dishes. After 18–24 hours, the medium was replaced with DMEM supplemented with 10% FBS

(without P/S). Cells were then transfected with 10 µg of transfer plasmid (e.g., shRNA

constructs), 8 µg of psPAX2 (Addgene #12260), and 2 µg of pMD2.G (Addgene #12259), using

X-tremeGENE 360 transfection reagent (Roche, 8724156001) at a 1:3 DNA-to-reagent ratio,

following the manufacturer's instructions. After 48–72 hours, the viral supernatant was collected,

centrifuged at 1,100 rpm for 5 minutes to remove cell debris, and filtered through a 0.45 µm PES

filter (Cytiva, 28143-312). The resulting virus-containing media were used to transduce target

cells in the presence of polybrene (MilliporeSigma, C788D57) at a final concentration of 1 µg/ml.

Twenty-four hours after transduction, cells were selected with 2 µg/ml puromycin (Thermo

Scientific, 53-79-2) for at least 72 hours. Successful gene knockdown was subsequently

validated by Western blot, confirming effective transduction and target suppression.

siNFYA transfection and JQ1 treatment

NFYA-specific siRNA (Santa Cruz Biotechnology, sc-29947) and non-silencing control siRNA

(Santa Cruz Biotechnology, sc-37007) were purchased from Santa Cruz Biotechnology. Cells

were transfected with 40 nM siRNA using X-tremeGENE 360 reagent (Roche, 8724156001)

following the manufacturer's protocol. After 48 hours, transfected cells were treated with JQ1 (1

µM for 24 hours), and NFYA knockdown efficiency was assessed by Western blot.

Western blotting

Cells at 80–90% confluency were scraped on ice using a cell lifter (Biologix, 70-2180) into lysis

buffer (50 mM Tris, pH 7.5; 150 mM NaCl; 5 mM EDTA; 1% Triton X-100) supplemented with

protease inhibitor cocktail (Sigma-Aldrich, SIAL-11836170001) and phosphatase inhibitor

cocktail (Roche, 04906837001). Lysates were incubated on ice for 30 minutes and then clarified

by centrifugation at 14,000 rpm for 10 minutes at 4°C. Supernatants were transferred to fresh

tubes, and protein concentrations were measured using the DC Protein Assay Kit (Bio-Rad;

Reagents A, B, S: 5000113, 5000114, 5000115). A total of 20 µg of protein was mixed with 4×

LDS sample buffer (Invitrogen, B0007) and 10× NuPAGE Sample Reducing Agent (Invitrogen,

NP0009), adjusted to a final volume of 20 µl, and heated at 70°C for 10 minutes. Samples were

loaded onto 4–12% Bis-Tris NuPAGE gels (Invitrogen, NP0321) and electrophoresed using a

Mini Gel Tank (Thermo Fisher, A25977) in MES SDS Running Buffer (Invitrogen, NP0002) at

120 V for 90 minutes. Proteins were transferred to 0.45 µm PVDF membranes (Millipore,

IPVH00010) at 350 mA for 150 minutes at 4°C. Membranes were blocked in 5% skim milk in

Tris-buffered saline (Bio-Rad, 1706435) with 0.1% Tween-20 (Fisher Scientific, BP337-500)

(TBST) for 1 hour at room temperature. They were then incubated overnight at 4°C with primary

antibodies (see Table S4) diluted in 5% skim milk in TBST under gentle agitation (~70 rpm)

(ONILAB, SK-O180-S). Following incubation, membranes were washed three times for

10 minutes each with 1× TBST on a rocker and then incubated with HRP-conjugated secondary

antibodies (see Table S4) in the blocking buffer for 1 hour at room temperature. Membranes

were subsequently washed three more times for 10 minutes with TBST. Signal was detected

using SuperSignal West Femto Maximum Sensitivity Substrate (Thermo Scientific, 34094) and

visualized using the Amersham Imager 600 (GE Healthcare Life Sciences). Knockdown

efficiency was quantified in ImageJ using the Gel Analysis tool by measuring band intensities

(area under the curve). Target protein levels were normalized to the loading control and

expressed relative to shScr for each cell line.

Animal studies and treatment regimen

Female athymic nude (*Nu/Nu*) mice (6 weeks old) were obtained from the Jackson Laboratory

(Strain #:002019) and housed under specific pathogen-free conditions in the Genome and

Biomedical Sciences Facility at the University of California, Davis. Mice were maintained with

controlled ambient temperature and a standard light–dark cycle. All procedures were approved

by the University of California Davis Institutional Animal Care and Use Committee (Protocol

#24016). JQ1 (MedChemExpress, HY-13030) was prepared by dissolving the drug in DMSO

(Sigma-Aldrich, 472301-500ML), then diluting in vehicle composed of 95% of a 20% SBE-β-CD

solution (MedChemExpress, HY-17031) in saline (Moltox, C790D04) to a final concentration of

5 mg/ml. For subcutaneous tumor engraftment, cells were trypsinized, passed through a 40 µm

cell strainer (Biologix, 15-1040), and resuspended in 50 µl of Cultrex Reduced Growth Factor

Basement Membrane Extract (Bio-Techne, 3533-010-02). Each mouse received 0.5×10^6 cells:

shScramble (shScr) control cells in the left flank and shBrd2 cells in the right flank. Three days

post-engraftment, mice were randomized into two treatment groups (n = 10 per group) and

administered either vehicle or JQ1 intraperitoneally at 50 mg/kg, once a day, five days per week.

Tumor size was measured using Vernier calipers three times per week, and tumor volume was

calculated as: $(\pi \times \text{length} \times \text{width}^2)/6$. At the end of the treatment period, mice were euthanized,

and tumors were harvested for histological analysis.

Histology and immunohistochemistry

Harvested tumors were fixed in 10% neutral buffered formalin (Fisher Scientific, 22-110-873) and

replaced with 70% ethanol (Thomas Scientific, C993Y81) on the following day. Formalin-fixed

paraffin-embedded sections with 5 µm thickness were prepared for immunohistochemistry (IHC),

which was performed at the Center for Genomic Pathology, University of California, Davis. For

Ki67 staining, antigen retrieval was performed for 45 minutes with citrate buffer at pH 6.0 in a

Decloaking Chamber (Biocare Medical) at 125°C and 15 psi. Slides were blocked with normal

goat serum then incubated with rabbit polyclonal Ki67 (NeoMarkers, RB1510P, 1:1000 dilution)

overnight at room temperature in a humidified chamber, followed by biotinylated goat anti-rabbit

secondary antibody (1:1000 dilution). The Vectastain® ABC elite kit and diaminobenzidine

peroxidase substrate kit (Vector Laboratories) were used for amplification and visualization of

signal, respectively Ki67 staining was quantified in ImageJ using the IHC Profiler plugin (15).

Nuclei classified as high positive or positive by the plugin were counted as Ki67-positive.

RNA extraction

Cells were lysed in 1 ml of TRIzol reagent (Thermo Fisher Scientific, 15596026) at room

temperature for 5 minutes with gentle agitation, then 200 µl of chloroform (Reagents, C755N59)

was added. Total RNA was purified using the PureLink RNA Mini Kit (Invitrogen, 12183018A)

according to the manufacturer's instructions. RNA concentration and purity were assessed using

a NanoDrop ND-1000 spectrophotometer (Thermo Scientific) prior to downstream applications.

Reverse transcription quantitative PCR (RT-qPCR)

Total RNA was extracted using TRIzol reagent (Invitrogen, 15596026). After 5 minutes of

incubation at room temperature, 200 µl of chloroform (Reagents, C755N59) was added, and

samples were vigorously mixed until homogeneous, followed by 5 minutes on ice. Samples were

centrifuged at 14,000 rpm for 15 minutes at 4°C, and the aqueous phase was transferred to a

new DNase/RNase-free tube. RNA was further purified using the PureLink RNA Mini Kit

(Invitrogen, 12183018A), including on-column DNase treatment (PureLink DNase Set,

Invitrogen, 12185010). RNA was eluted in nuclease-free water, and concentration and purity

were assessed using a NanoDrop ND-1000 spectrophotometer (Thermo Scientific).

cDNA synthesis was performed with the High-Capacity cDNA Reverse Transcription Kit (Applied

Biosystems, 4368814) using 1 µg of total RNA per 20 µl reaction. The reaction mix included 3 µl

10× RT buffer, 1.2 µl 25× dNTP mix, 3 µl 10× random primers, 1.5 µl MultiScribe reverse

transcriptase, and 6.3 µl nuclease-free water. The thermocycling conditions were: 25°C for 10

minutes, 37°C for 120 minutes, and 85°C for 5 minutes.

qPCR was performed using 1 µl of cDNA with PowerTrack SYBR Green Master Mix (Applied

Biosystems, A46109) and gene-specific primers (listed in Table S5) in a total reaction volume of

10 µl. Reactions were run on a LightCycler 480 system (Roche) with the following setting: 95°C

for 10 minutes, followed by 40 cycles of 95°C for 15 seconds and 60°C for 60 seconds. Relative

mRNA expression levels were normalized to *GAPDH* (human samples) or *Gapdh* (mouse

samples) and calculated using the $2^{-\Delta\Delta Ct}$ method.

Chromatin immunoprecipitation assays (ChIP-seq and ChIP-qPCR)

BRD2 and BRD4 ChIP-seq ChIP

was performed using the SimpleChIP® Enzymatic Chromatin IP Kit (Magnetic Beads) (Cell

Signaling Technology, 9003) with minor modifications. Briefly, approximately 1×10^7 mT3-2D or

mT23-2D cells were dissociated with TrypLE Express (Gibco, 12605028), resuspended in PBS,

crosslinked with 1% formaldehyde for 10 minutes at room temperature, and quenched with 125

mM glycine for 5 minutes. Chromatin was digested with MNase (1 µL, 37°C, 20 minutes) and

sonicated (Bioruptor, Diagenode) at a high power (3 rounds of 10 cycles, 30 seconds on/30

seconds off) to obtain ~150–900 bp fragments. A 2% input aliquot (10 µl) was reserved, and the

remaining chromatin was incubated overnight at 4°C with constant rotation with 10 µl of BRD2

(Cell Signaling Technology, 5848S) or BRD4 (Bethyl, A301-985A100) antibody per reaction.

Bead capture, washes, elution, reverse crosslinking, and DNA purification were performed

according to the kit protocol.

<u>NFYA ChIP-qPCR</u>

ChIP-qPCR was performed similarly using approximately 4×10^6 HEK293T cells per reaction,

except chromatin was digested with 0.5 µl MNase (37°C, 10 minutes) and sonicated for 6 cycles

(30 seconds on/30 seconds off at high power). Immunoprecipitations used 1 µl of NFYA antibody

(Diagenode, C15310261) or 10 µl IgG control (Cell Signaling Technology, 2729S). qPCR was

performed using PowerTrack SYBR Green Master Mix (Applied Biosystems, A46109) with

primers in Table S5, following the qPCR conditions described in the RT-qPCR section. Percent

input was calculated as $100 \times 2^{(Ct\ adjusted\ input\ -\ Ct\ ChIP)}$. ChIP-seq and ChIP-qPCR were generated

using a single biological replicate per condition.

Library preparation and sequencing

Library preparation and sequencing were carried out by Novogene Co., LTD. For RNA-seq, total

RNA quality was assessed using 1% agarose gel electrophoresis, NanoDrop spectrophotometry,

and an Agilent 2100 Bioanalyzer. Poly(A) mRNA was enriched from total RNA using oligo(dT)

magnetic beads, fragmented, and reverse-transcribed to synthesize first-strand cDNA with

random primers. Second-strand cDNA synthesis was performed using a dUTP-incorporation

protocol to maintain strand specificity. The resulting double-stranded cDNA was end-repaired, A-

tailed, and ligated to Illumina adapters, followed by size selection (250–300 bp insert) and PCR

amplification. Libraries were quantified, quality-checked, and sequenced on an Illumina

NovaSeq X Plus platform to generate paired-end 150 bp reads (PE150). For ChIP-seq, sample

quality was assessed prior to library construction. DNA fragments were end-repaired, A-tailed,

and ligated to Illumina adapters, followed by size selection, PCR amplification, and quantification.

Libraries were sequenced on an Illumina NovaSeq X Plus platform (PE150). Quality control

procedures, including assessments of sequencing quality distribution, error rate distribution, and

AT/CG base composition, were performed before data delivery.

Bioinformatic analysis

All analyses were performed on the Galaxy platform unless otherwise stated. Raw reads were

quality-trimmed with Trim Galore! to remove adapter sequences and low-quality bases.

RNA-seq analysis

For RNA-seq, clean reads were aligned to the *Mus musculus* mm9 reference genome using

HISAT2, and gene-level counts were obtained with featureCounts. Normalization and differential

expression analysis were performed with DESeq2, using the median-of-ratios method. P-values

were adjusted for multiple testing using the Benjamini–Hochberg procedure, and genes with

adjusted $p < 0.05$ were considered significantly differentially expressed. Normalized counts were

obtained from DESeq2, annotated using annotatemyIDs, and used for downstream visualization.

Volcano plots were generated by plotting the log_2FC against $-log_{10}p$-adj to highlight significantly

upregulated and downregulated genes.

ChIP-seq analysis

All ChIP-seq samples were processed using the same analysis pipeline with identical parameters

as described below. Trimmed reads were aligned to the *Mus musculu*s mm9 reference genome

with Bowtie2, then filtered with SAMtools to retain only properly paired reads with a mapping

quality ≥ 30 while excluding mitochondrial reads (chrM). PCR and optical duplicates were

removed with Picard MarkDuplicates. To ensure that only high-quality, non-redundant reads

were used in downstream analyses, we evaluated ChIP enrichment using standard QC metrics.

Across all BRD2/BRD4 ChIP-seq samples, the mean mapped read depth was 17.4 million

reads. Enrichment assessed by FRiP averaged 0.33 across IP samples, and IP enrichment

relative to input was additionally supported by deepTools plotFingerprint.

Peaks were called with MACS2. Consensus peaks were defined as those present in both

biological replicates using bedtools intersect. For each cell line, BRD2/BRD4 co-occupied

regions were identified by intersecting BRD2 and BRD4 consensus peak sets. Regions were

recentered on peak summits and extended ± 2 kb to create fixed windows for signal extraction.

Signal coverage tracks (BigWig files) were generated from duplicate-removed BAM files using

bamCoverage with RPGC normalization (scaled to scaled to 1× genome coverage). Signal

changes across ranked regions were visualized with deepTools (plotHeatmap, plotProfile). Per-

region signal was extracted using deepTools computeMatrix reference-point (±2,000 bp, 50-bp

bins, refPoint = center, skipZeros = true, missingDataAsZero = true). For each region, bins were

averaged to obtain a single signal value per replicate and condition (DMSO, JQ1). Using a single

ChIP-seq dataset per condition, we used a rank-based, region-level analysis to summarize

relative signal changes and interpreted these results as descriptive.

To define BRD2 regions most sensitive to BET inhibition in the sensitive cell line, a per-region

change in mT3-2D was calculated as $\Delta = \text{BRD2}_{DMSO} - \text{BRD2}_{JQ1}$, where larger positive Δ indicates

greater BRD2 loss following JQ1. The top 5% mT3 JQ1-sensitive BRD2 regions (mm9

coordinates) were assigned to candidate genes using GREAT with the default basal-plus-

extension model (proximal: 5 kb upstream and 1 kb downstream of the transcription start site;

distal up to 1 Mb), yielding 1,035 associated genes. Functional enrichment for GO Biological

Process (GOBP) terms was first assessed using GREAT, and the same gene set was additionally

analyzed using ShinyGO GOBP (default mouse genome background; Benjamini–Hochberg FDR

correction; q < 0.05) for complementary visualization and reporting. The top 10 enriched terms

were reported. For integrative expression analysis, GREAT-associated genes were intersected

with the RNA-seq expression matrix (DESeq2-normalized counts), resulting in 1,028 genes that

were visualized in Morpheus as heatmaps of row-scaled z-scores (z-score computed per gene

across samples).

Analysis of publicly available datasets

<u>RNA-seq analysis of BRD2 response to BET inhibition (JQ1)</u>

RNA-seq expression data from human conventional cancer cell lines were obtained via the

GEO2R portal. Differential expression analysis was performed using the limma package as

implemented in GEO2R. Log_2FC and $-log_{10}p$-adj for *BRD2* were extracted to evaluate its response

to JQ1 treatment, and these values were visualized as dot plots in R. To visualize differential

expression patterns after JQ1 treatment, 3D volcano plots were generated in R (version 4.3.0)

using DESeq2 output from GEO2R, with specific focus on *BRD2* and *NFYA*. A summary of the

datasets analyzed, including GEO accession numbers, cancer types, and sample details, is

provided in the Table S6.

<u>RNA-seq analysis of BRD4 knockdown</u>

RNA-seq data from human conventional cancer cell lines were obtained from the GEO2R portal.

Normalized counts were used as provided, and BRD2 expression was compared between control

and BRD4 knockdown conditions. The datasets used in this analysis are listed in Table S7.

<u>ChIP-seq analysis of BRD2, BRD4, and NFYA</u>

BRD2 and BRD4 ChIP-seq data from human breast cancer cell lines (GSE131097) were obtained

from GEO (listed in Table S8). Processed peak files (BED) and signal tracks (BigWig) were used

directly from the repository. For BRD2 and BRD4 ChIP-seq, binding sites were intersected using

bedtools intersect to define co-occupied regions, and signal distribution across these regions was

assessed with computeMatrix followed by plotProfile to generate metaprofiles. For NFYA,

H3K4me3, RNA Pol II ChIP-seq, browser tracks of each BigWig file from GEO (listed in Table S9)

were visualized in IGV (version 2.13.2) at the *BRD2* locus.

Statistical analysis

All statistical analyses were performed using GraphPad Prism version 10.6.0 (796). Statistical

significance between two groups was assessed using a two-tailed, unpaired Student's t-test. For

comparisons involving more than two groups, one-way ANOVA with Tukey's post hoc test was

used. A $p < 0.05$ was considered statistically significant. Error bars represent the SEM unless

otherwise indicated.

Results

BET inhibition upregulates BRD2 across cancer types

To identify adaptive transcriptional responses to BET inhibition across cancer types, we first performed a systematic analysis of publicly available transcriptomic data. In total, we curated 51 human RNA-seq datasets following treatment with a small molecule BETi JQ1 from the Gene Expression Omnibus (GEO), published over the past decade (2014–2024), encompassing a wide spectrum of malignancies including prostate, breast, brain, skin, gastrointestinal, lung, pediatric, gynecologic and hematologic cancers (16-64). Across all datasets analyzed, *BRD2* expression was consistently and significantly upregulated upon JQ1 treatment, emerging as a recurrent transcriptional response to BET inhibition **(Fig. 1A)**, while no consistent induction was observed for BRD3 and BRD4. The strongest increase in *BRD2* expression was observed in castration-resistant prostate cancer (CRPC; GSE78213), where JQ1 treatment induced a significant upregulation ($\log_2$FC = 2.15; p-adj = 0.00), whereas *BRD3* and *BRD4* exhibited only modest changes ($\log_2$FC = -0.13, p-adj = 0.0931; $\log_2$FC = 0.15, p-adj = 0.561, respectively) **(Fig. 1B)**. To validate the findings from the publicly available datasets, we selected cell lines from multiple cancer types represented in the GEO analysis and assessed *BRD2* expression after JQ1 treatment. These included CRPC (PC3), hepatocellular carcinoma (HCC; Huh7), triple-negative breast cancer (TNBC; MDA-MB-231 and SUM159), glioblastoma (GBM; U251), neuroblastoma (NB; SK-N-BE2), non-small cell lung cancer (NSCLC; A549), osteosarcoma (OS; U2OS), mouse and human acute myeloid leukemia (AML; RN2 and HL-60). In all cell lines tested, BRD2 protein levels were significantly increased following JQ1 treatment **(Fig. 1C; Fig. S1A)**. Notably, BRD2 expression was generally low in leukemia cell lines that we tested. In human AML (HL-60), BRD2 was undetectable at the protein level under baseline conditions; however, we confirmed its induction at the mRNA level in response to JQ1. Similarly, in mouse AML (RN2), BRD2 protein became detectable only after JQ1 exposure **(Fig. S1B–S1C)**. Collectively, these findings

confirmed that BRD2 upregulation emerged as a universal response to BET inhibition,

distinguishing it from other BET family members. These results suggest a previously

underappreciated role for BRD2 in the context of pharmacological BET blockade and raise the

possibility that its upregulation reflects a feedback or compensatory mechanism.

Pharmacological BET inhibition, not BRD4 depletion, induces BRD2 upregulation in

pancreatic cancer

Previously, we reported that BRCA2 deficiency sensitized pancreatic ductal adenocarcinoma

(PDAC) cells to BETi (65), in contrast to the general resistance observed in most PDAC cells to

BETi. To further investigate the underlying mechanisms of *BRD2* upregulation upon BET

inhibition, we used PDAC as a model system. In the publicly available dataset GSE192903,

which profiles PANC-1 cells treated with JQ1, BRD2 was significantly upregulated ($\log_2$FC =

1.18, p-adj = 4.87×10^{-185}) **(Fig. 2A)**. To validate this, we treated 3 murine KPC pancreatic

cancer cell lines (mT3-2D, mT19-2D, and mT23-2D) with JQ1 and conducted RNA-seq. Principal

component analysis (PCA) revealed distinct transcriptomic shifts upon the treatment, indicating

that JQ1 broadly altered the transcriptional landscape **(Fig. S2A)**. In line with their transcriptional

activation role, JQ1 treatment led to widespread gene downregulation, with 1,839 genes

significantly decreased. Although *Brd3* and *Brd4* expression remained unchanged, 353 genes

were significantly upregulated. Notably, *Brd2* was among the most significantly induced genes

($\log_2$FC = 2.25, p-adj = 7.16×10^{-25}) **(Fig. 2B)**. This finding was further validated by RT-qPCR

and Western blot in both murine primary and metastatic PDAC cells, along with a panel of

human PDAC cell lines, confirming that BRD2 upregulation is a robust and reproducible

transcriptional response to BET inhibition across species and disease stages **(Fig. 2C–D; Fig.**

S2B–D). Importantly, this effect was not limited to JQ1; treatment with other small-molecule

BETi, including birabresib, molibresib, CPI-0610, ZEN-3694, and AZD5153, also elevated BRD2

protein expression **(Fig. 2E; Fig. S2E–F)**. Because BRD4 is a major BET family with well-

established role in transcriptional regulation, we next asked whether BRD2 upregulation could

be induced by genetic depletion of BRD4. To test this, we knocked down BRD4 using shRNA in

the same murine PDAC cell lines. Despite successfully silencing BRD4, BRD2 expression

remained unchanged, highlighting that BRD4 KD elicits a different transcriptional response from

pharmacological BET inhibition. **(Fig. 2F; Fig. S2G)**. Consistent with this observation, BRD2

was upregulated by JQ1 in various cancer models, but not by genetic depletion of BRD4 **(Fig.**

S2H) (47, 55, 66-69). While the transcriptomic profile of shBRD4 cells shifted in a similar

direction to that of JQ1-treated cells, the changes were less pronounced, indicating that BRD4

KD only partially phenocopies BET inhibition **(Fig. 2G)**. Together, these findings demonstrate

that pharmacological BET inhibition induces BRD2 upregulation as part of a broader

transcriptional response that is mechanistically distinct from BRD4 depletion, potentially

reflecting an adaptive program to preserve transcriptional output.

NFYA is one of the potential upstream regulators of BRD2 in response to BET inhibition

BETi-mediated induction of BRD2 prompted us to investigate the upstream factors that might be

linked to its regulation. Given that gene induction is often shaped by transcription factors (TF),

we sought to nominate TF candidates that might regulate BRD2. We began by querying the

COEXPRESdb platform for transcription factors co-expressed with BRD2. Among the top 30 co-

expressed genes, Nuclear Transcription Factor Y subunit Alpha (NFYA) was the only TF

identified **(Fig. S3A–B)**. Consistent with this, our analysis of transcriptomic data from DepMap

across 1,684 human cancer cell lines revealed a positive correlation between *NFYA* and *BRD2*

expression (Pearson's *r* = 0.628, p = 0.00) **(Fig. 3A)**. We next asked whether *NFYA* and *BRD2*

exhibit coordinated responses to BET inhibition. Analysis of both publicly available RNA-seq

datasets and our mouse PDAC dataset revealed that both *NFYA* and *BRD2* were upregulated

under JQ1 treatment **(Fig. 3B; Fig. S3C)**. To evaluate whether NFYA acts as an upstream

regulator of *BRD2* transcription, we analyzed public NFYA ChIP-seq datasets and observed

NFYA signal near the *BRD2* transcription start site (TSS) within an active promoter region

marked by H3K4me3 and RNA Pol II occupancy across multiple cell types **(Fig. 3C; Fig. S3D)**.

To test this relationship, we performed siRNA-mediated KD of NFYA in HEK293T and U2OS

cells. Depletion of NFYA reduced basal levels of *BRD2* expression and partially reduced BRD2

expression after JQ1 treatment **(Fig. 3D–G; Fig. S3E–F)**. Indeed, NFYA ChIP-qPCR showed

the increased NFYA binding at the BRD2 promoter region upon JQ1 treatment relative to IgG

control **(Fig. S3D)**, suggesting that NFYA may contribute, at least in part, to *BRD2* transcriptional

upregulation in response to BET inhibition.

BRD2 knockdown sensitizes cells to BET inhibition *in vitro*

Given the robust upregulation of BRD2 following BET inhibition, we hypothesized that BRD2

upregulation might confer adaptive resistance to BETi. To test this, we performed shRNA-

mediated KD of BRD2 in a panel of cancer cell lines and evaluated their sensitivity to BETi. In

mouse PDAC cells, BRD2 KD enhanced sensitivity to 6 different BETi—JQ1, birabresib,

molibresib, ZEN-3694, CPI-0610, and AZD5153 **(Fig. 4A–B)**. Similar results were also observed

in 3 additional human cancer cell lines—A549 (NSCLC), U251 (GBM), and U2OS (OS)—where

BRD2 KD consistently increased JQ1 sensitivity **(Fig. 4C–H)**. Across multiple cell lines, BRD2

depletion sensitized cells to BET inhibition, supporting that BRD2 upregulation contributes to

adaptive resistance against BET inhibition **(Fig. S4A–I)**. Moreover, clonogenic assays under

chronic treatment mirrored this response, with durable suppression of colony outgrowth over

extended BETi exposure **(Fig. S4J–O)**. Interestingly, we observed that the magnitude of

sensitization varied depending on the cell's initial response to BETi. Cells that were more

resistant to BET inhibition at baseline (mT23-2D, mM1-2D, mM6-2D, and Capan-2) exhibited a

greater reduction in IC_{50} upon BRD2 KD compared to those that were initially sensitive (mT3-2D,

CFPAC-1, and BxPC3) **(Fig. 4I–N, Fig. S4A–I)**. In sum, our results revealed a functional role for

BRD2 in mediating an acute adaptive response to BETi, suggesting that targeting BRD2

enhances therapeutic efficacy, particularly in BETi-resistant tumors.

BRD2 depletion enhances BETi efficacy *in vivo*

To evaluate whether BRD2 contributes to BETi response *in vivo*, we subcutaneously injected

mouse pancreatic cancer cells (mT23-2D) harboring control or BRD2-targeting shRNA into

immunocompromised athymic *Nu/Nu* mice. Mice were treated intraperitoneally with either

vehicle or JQ1 (50 mg/kg) over a three-week period **(Fig. 5A)**. While both JQ1 treatment alone

or BRD2 KD alone moderately suppressed tumor growth, combined BRD2 KD and JQ1

treatment resulted in striking tumor growth suppression **(Fig. 5B–D)**. No overt toxicity was

observed as body weight remained stable across treatment arms **(Fig. S5)**. Immunohistochemical

analysis of Ki67 staining revealed a marked reduction in proliferative cells in BRD2 KD tumors

treated with JQ1, which exhibited the lowest percentage of Ki67-positive cells **(Fig. 5E–F)**.

Together, these results demonstrate that BRD2 depletion significantly enhances the antitumor

efficacy of BET inhibition *in vivo*.

BRD2 persists on chromatin preferentially in BETi-resistant cells

Given our findings that BRD2 is associated with adaptive resistance to BET inhibition both *in vitro*

and *in vivo*, we sought to investigate potential underlying mechanisms. We first examined the

basal expression levels of BRD4 and BRD2 but did not observe a clear correlation with JQ1 IC_{50}

values **(Fig. S6A–B)**, consistent with DepMap data across 394 cell lines (Pearson's *r* = -0.102

and -0.058 for BRD4 and BRD2, respectively) **(Fig. S6C–D)**. Since BETi displaces BET proteins

from acetylated chromatin, we hypothesized that differential BETi responses between BETi-

sensitive (mT3-2D) and BETi-resistant (mT23-2D) cells **(Fig. 6A; Fig. S6E)** may reflect differential

BRD2 and BRD4 chromatin occupancy after BETi treatment. To address this, we performed BRD2

and BRD4 ChIP-seq in both BETi-sensitive (mT3-2D) and BETi-resistant (mT23-2D) cells to

compare their chromatin occupancy under DMSO and JQ1 conditions. In the DMSO condition,

we identified common BRD4 (n = 24,810) and BRD2 peaks (n = 23,706) between these two cell

lines. Notably, about 70% of BRD4-bound regions (16,941 of 24,810 peaks) were also bound by

BRD2 **(Fig. 6B)**. In these overlapping regions, upon JQ1 treatment, we found that BRD4

chromatin occupancy overall decreased in both cell lines to a similar extent. However, BRD2

occupancy was retained to a greater degree in BETi-resistant (mT23-2D) cells than BETi-sensitive

(mT3-2D) cells **(Fig. 6C)**. A similar pattern of BRD2 persistence has been observed in TNBC cell

line (SUM149) after BET inhibition, suggesting a broader relevance across cancer types **(Fig.**

S6F–G). To further examine whether differential BRD2 displacement affects transcriptional output,

we focused on the top 5% of BRD2/BRD4 co-occupied sites showing the greatest loss of BRD2

binding after JQ1 treatment in mT3-2D cells (847 regions), which were comparatively better

preserved in JQ1-treated mT23-2D cells. Consistent with this binding pattern, our RNA-seq

analysis revealed a more pronounced transcriptional shift for genes associated with these sites

in mT3-2D cells, whereas the corresponding transcriptional profiles in mT23-2D cells remained

closer to DMSO-treated controls **(Fig. 6D)**. Gene ontology analysis of this gene set showed

enrichment for chromatin organization/modification and RNA processing programs, suggesting

potential roles of these genes in mediating adaptive resistance of BETi **(Fig. 6E; Fig. S6H)**.

Collectively, our data support a model in which, upon BET inhibition, resistant behavior is

associated with greater BRD2 chromatin retention and more preserved transcriptional output

despite reduced BRD4 occupancy **(Fig. 6F)**. These findings suggest that BRD2 may serve as a

compensatory factor when BRD4 is displaced from chromatin and provide a rationale for targeting

BRD2 in combination with BETi to improve therapeutic responses.

Discussion

BRD4 dependency in many cancers, including leukemia, provided the rationale for developing

BETi. The first-generation inhibitor JQ1 demonstrated potent antitumor effects in preclinical

models such as NUT carcinoma and MYC-driven leukemia. However, its poor bioavailability,

short half-life, and off-target toxicities limited its clinical use, prompting the development of

second-generation BETi with improved pharmacokinetics, enhanced specificity, and orally

available formulations. Indisputably, despite these advances, cancer cells can activate

alternative pathways to circumvent therapy, mitigating its effects and sustaining survival. These

adaptive responses not only represent tangible vulnerabilities but, when understood, can also

be exploited to prevent resistance and achieve better therapeutic outcomes. Several studies

have informally noted increased BRD2 expression following BETi treatment, but it has remained

unclear whether this functionally confers pan-cancer adaptive resistance. Recent synthetic lethal

screens with BETi in TNBC and paralog-focused CRISPR screens in lung, skin, pancreas, and

brain cancers have independently identified BRD2 as a potential resistance factor (70, 71). Yet,

the role of BRD2 has not been systematically examined. Here, we show that transcriptional

upregulation of BRD2 is a robust adaptive response to BETi across diverse cancer types, and

that BRD2 depletion sensitizes cells to BET inhibition both *in vitro* and *in vivo*.

Paralog compensation is an emerging theme in targeted cancer therapy, wherein inhibition of

one protein prompts adaptive upregulation of a closely related family member that compensates

for its function. For example, EZH2 inhibition in malignant rhabdoid tumors has been shown to

induce EZH1 upregulation, and dual EZH1/2 inhibition yields improved therapeutic responses

(72). Similarly, inhibition of the histone demethylase LSD1 (KDM1A) in Ewing sarcoma cells

resulted in marked induction of its homolog LSD2 (KDM1B), a shift that modulates drug

sensitivity (73). Recent large-scale paralog-focused CRISPR screens have uncovered multiple

additional examples of this phenomenon, such as the reciprocal buffering between DUSP4 and

DUSP6 (71). Here, we present the first comprehensive evidence that BRD2 upregulation is a

highly conserved and universal adaptive response to BET inhibition across cancer types,

functionally analogous to other paralog compensation events in which BRD2 substitutes for

BRD4 to preserve transcriptional programs upon BET inhibition.

Our data suggest that resistance is mediated in part through paralog redundancy, with BRD2

compensating for BRD4 function under BET inhibition. This compensatory role aligns with their

shared canonical functions in transcriptional regulation, as evidenced by their largely overlapping

genomic occupancy (70). However, BRD2 also carries out distinct functions that may

independently and uniquely support adaptive resistance. Upon JQ1 treatment, BET proteins can

remain in the nucleus and, through their extra-terminal (ET) domains, may exert bromodomain-

independent functions that sustain cell survival. Unlike BRD4, BRD2 is closely associated with

chromatin architectural regulation: it co-localizes genome-wide with CTCF, contributes to the

maintenance of topologically associating domain (TAD) boundaries, and influences

compartmentalization of the accessible genome (74, 75). The C-terminal domain of BRD2 also

mediates critical protein-protein interactions, enabling its recruitment to promoters and CTCF-

bound sites. The unique roles of BRD2 raise the possibility that its upregulation under BET

inhibition may confer survival advantages beyond its role in transcriptional regulation.

Our findings suggest a therapeutic opportunity to enhance BETi efficacy by preventing or

counteracting BRD2 upregulation. This could be achieved either through selective BRD2

inhibition in combination with conventional BETi which predominantly target BRD4 or by

strategies that broadly degrade BET family members to prevent compensatory mechanisms.

PROTAC-based BET degraders, which remove BRD2, BRD3, and BRD4 from chromatin, have

recently demonstrated superior antitumor activity compared to earlier generations of BETi (59,

76). Notably, the absence of BRD2 upregulation in PROTAC-based approaches may partially

explain their improved efficacy. While additional resistance mechanisms undoubtedly exist, our

data highlight BRD2 compensation as a conserved and targetable pathway. As a proof-of-

principle, we show that BRD2 depletion sensitizes tumors to BET inhibition. This provides a

rationale for integrating BETi with approaches that mitigate BRD2-driven adaptive resistance to

achieve more durable clinical responses and expand the therapeutic potential of BET-targeted

strategies in cancer therapeutics.

 Figure legends

Fig. 1 Pan-cancer BRD2 upregulation upon BET inhibition. (A) Dot plots showing *BRD2*

expression changes (JQ1 vs. control) across 51 curated human RNA-seq datasets from GEO

(2014–2024), including gastrointestinal, breast, prostate, skin, brain, lung, pediatric, gynecologic,

and hematologic cancers. The x-axis represents $\log_2$fold change ($\log_2$FC), and dot size

corresponds to statistical significance ($-\log_{10}$p-adj). Abbreviations: CRPC, castration-resistant

prostate cancer; TNBC, triple-negative breast cancer; Lum A, luminal A; BC, breast cancer;

GBM, glioblastoma; NB, neuroblastoma; MEL, melanoma; HNSCC, head and neck squamous

cell carcinoma; CRC, colorectal cancer; PDAC, pancreatic ductal adenocarcinoma; GIST,

gastrointestinal stromal tumor; HCC, hepatocellular carcinoma; SCLC, small cell lung cancer;

NSCLC, non-small cell lung cancer; RMS, rhabdomyosarcoma; AT/RT, atypical teratoid/rhabdoid

tumor; ES, Ewing sarcoma; OC, ovarian cancer; CC, cervical cancer; MM, multiple myeloma; T-

ALL, T-cell acute lymphoblastic leukemia; CML, chronic myeloid leukemia; BL, Burkitt lymphoma;

AML, acute myeloid leukemia. **(B)** Volcano plot of differentially expressed genes in CRPC

(GSE78213), highlighting robust *BRD2* upregulation upon JQ1 treatment. *BRD3* and *BRD4*

remained unaffected. **(C)** Western blot validation showing BRD2 protein levels in multiple

representative human cell lines from prostate (PC3), liver (Huh7), breast (MDA-MB-231), brain

(U251), lung (A549), and pediatric (U2OS) cancers with and without JQ1.

Fig. 2 Pharmacological BET inhibition uniquely drives BRD2 upregulation in pancreatic

cancer. (A) Volcano plot of differentially expressed genes in human PDAC (GSE192903) treated

with JQ1, highlighting strong upregulation of *BRD2*, neither *BRD3* nor *BRD4*. **(B)** Volcano plot of

RNA-seq data from murine PDAC cells treated with JQ1, showing conserved and significant

upregulation of *Brd2*. **(C)** Western blot validation of BRD2 upregulation in murine PDAC cell

lines treated with JQ1. **(D)** Western blot analysis in a panel of human PDAC cell lines confirming

JQ1-induced BRD2 expression. **(E)** Western blot analysis of BRD2 in murine PDAC cells (mT23-

2D) treated with different small-molecule BET inhibitors, showing a consistent increase in BRD2

expression across inhibitors. **(F)** Western blot analysis of BRD2 expression following shRNA-

mediated BRD4 knockdown (KD) in murine and human PDAC cell lines, revealing that BRD4

depletion does not mimic JQ1-induced BRD2 upregulation. **(G)** Principal component analysis

(PCA) of RNA-seq data from murine PDAC cells under DMSO, JQ1, shScramble (shScr), or

shBrd4 conditions, showing distinct transcriptional clustering of JQ1-treated samples compared

to BRD4 KD.

Fig. 3 NFYA as a potential candidate transcriptional factor associated with BRD2 induction

upon BET inhibition. (A) Scatter plot from DepMap analysis showing a positive correlation

between *NFYA* and *BRD2* expression across public cancer transcriptomes. **(B)** Representative

3D volcano plot of publicly available RNA-seq datasets treated with JQ1, highlighting co-

upregulation of *BRD2* and *NFYA*. **(C)** NFYA ChIP-seq profiles at the *BRD2* locus across multiple

cell types. **(D,E)** Western blot analysis of NFYA and BRD2 in HEK293T (D) and U2OS (E) cells

transfected with control siRNA (siNeg) or NFYA siRNA (siNFYA) and treated with DMSO or JQ1.

(F,G) RT-qPCR analysis of *BRD2* expression in HEK293T (F) and U2OS (G) cells under the same

conditions, showing that NFYA KD attenuated JQ1-mediated *BRD2* upregulation. Statistical

significance was determined by one-way ANOVA ($*p < 0.05$, $**p < 0.01$, $***p < 0.001$, $****p <$

0.0001).

Fig. 4 BRD2 depletion sensitizes cancer cells to BET inhibition. (A) Immunoblot confirming

BRD2 KD (shBRD2 #1) in mT23-2D PDAC cells. Values indicate relative BRD2 protein levels

quantified by densitometry, normalized to GAPDH, and expressed relative to shScr. **(B)** Dose–

response curves showing reduced cell viability upon treatment with the indicated BETi in BRD2-

depleted mT23-2D cells compared to shScr controls. **(C–N)** Immunoblots confirming BRD2 KD

(shBRD2 #1) and JQ1 dose–response curves in lung (A549), brain (U251), pediatric (U2OS), and

human PDAC (Capan-2, CFPAC-1, and BxPC-3) cell lines following BRD2 KD.

Fig. 5 Genetic BRD2 perturbation enhances the antitumor effect of BET inhibition *in vivo*.

(A) Schematic of *in vivo* experimental design, created with BioRender. **(B)** Representative images of mice and excised tumors from each group after treatment. Scale bar, 1 cm. **(C)** Tumor growth curves over time showing significantly impaired tumor growth in shBrd2 + JQ1 tumors compared to controls. Error bars represent SD. **(D)** Tumor weight at endpoint, showing that shBrd2 + JQ1 tumors had the lowest weights. Statistical significance was determined by two-tailed Student's t-test (****$p < 0.0001$). **(E–F)** Representative Ki67 immunohistochemistry and quantification showing significantly reduced proliferative index in shBrd2 + JQ1 tumors compared to controls. Statistical significance was determined using two-tailed Student's t-test (****$p < 0.0001$). Scale bar, 100 μm.

Fig. 6 BRD2 chromatin retention may help maintain transcriptional programs in BETi-resistant pancreatic cancer cells. (A) Dose–response curves comparing sensitivity to BET inhibitors in BETi-sensitive (mT3-2D) and JQ1-resistant (mT23-2D) PDAC cells. **(B)** Venn diagram showing overlap between BRD4 and BRD2 ChIP-seq peaks across BETi-sensitive (mT3-2D) and BETi-resistant (mT23-2D) cells. **(C)** Metaplots and heatmaps of BRD4 and BRD2 ChIP-seq signal at BRD2/BRD4-overlapping peaks (16,941 regions) in BETi-sensitive (mT3-2D) and BETi-resistant (mT23-2D) cells treated with DMSO or JQ1. **(D)** RNA-seq heatmap (row-scaled z-scores) and BRD2 ChIP-seq browser tracks at representative loci in both BETi-sensitive (mT3-2D) and BETi-resistant (mT23-2D) cells **(E)** GREAT gene ontology enrichment for genes associated with top 5% BRD2-loss sites after JQ1 in mT3-2D cells. **(F)** Proposed schematic model illustrating BRD2 upregulation as an adaptive mechanism to BET inhibition.

Supplementary figure legends

Supplementary Fig. 1 BRD2 upregulation by JQ1 across additional cancer types. (A) Western blot analysis of BRD2 in brain and breast cancer cells treated with DMSO or JQ1. **(B–C)**

RT-qPCR and immunoblot analysis of BRD2 expression in human and mouse blood cancer cells,

showing increased BRD2 levels upon JQ1 treatment (1 µM, 24 hours).

Supplementary Fig. 2 BRD2 induction by BET inhibition but not by BRD4 KD. (A) PCA of

RNA-seq profiles from three different murine PDAC cell lines treated with DMSO or JQ1, showing

distinct clustering of JQ1-treated samples. **(B)** RT-qPCR analysis of *BRD2* expression in multiple

murine and human PDAC cell lines treated with JQ1, revealing significant induction across

models. Statistical significance was determined using a two-tailed Student's t-test (***p < 0.001,

****p < 0.0001) **(C–D)** Western blot analysis of BRD2 and BRD4 expression in murine (mT3-2D,

mM1-2D) and human (BxPC-3, Capan-1) PDAC cells after JQ1 treatment. **(E)** Immunoblot

analysis of BRD2 and BRD4 expression following treatment with JQ1, Birabresib or Molibresib in

mouse PDAC cells (mT3-2D and mT19-2D). **(F)** Western blot analysis of BRD2 expression in

murine PDAC cell lines (mT4-2D, mT5-2D, and mT8-2D) treated with different BETi (JQ1,

Birabresib, Molibresib, ZEN-3694, CPI-0610, AZD5153), showing consistent BRD2 upregulation

across inhibitors compared to control. **(G)** Western blot analysis of BRD2 expression after shRNA-

mediated BRD4 KD in PDAC AsPC-1 and breast cancer SUM159 cells, showing that BRD4

depletion did not result in BRD2 induction by JQ1. **(H)** Analysis of publicly available RNA-seq

datasets (GEO) across multiple cancer types, showing that BRD4 KD generally did not increase

BRD2 expression, whereas JQ1 significantly induced *BRD2*. Statistical significance was

determined by two-tailed Student's t-test (**p < 0.01).

Supplementary Fig. 3 NFYA emerges as a candidate regulator of BRD2. (A) List of top 30

BRD2-interacting transcriptional regulators identified from COEXPRESdb. **(B)** Co-expression

analysis of NFYA and BRD2 using COEXPRESdb, showing a strong positive correlation across

transcriptomes (z = 5.62). **(C)** RNA-seq analysis of *Nfya* expression in mouse PDAC cells treated

with DMSO or JQ1, showing significant *Nfya* upregulation upon JQ1 treatment. Statistical

significance was determined by two-tailed Student's t-test (***p < 0.001). **(D)** Public ChIP-seq

tracks showing H3K4me3, RNA pol II, and NFYA binding across multiple cell types at the BRD2

locus. Lower panels: NFYA ChIP-qPCR in HEK293T cells under DMSO or JQ1 treatment showing

enrichment at the *BRD2* exon 1 region relative to IgG and an intergenic negative control.

Statistical significance was determined by two-tailed Student's t-test (**$p < 0.01$). **(E–F)** Western

blot analyses showing NFYA and BRD2 expression in control (shScr) and NFYA- KD (shNfya)

murine PDAC lines mT5-2D (E) and mT23-2D (F) following DMSO or JQ1 treatment. NFYA

depletion reduces BRD2 induction upon BET inhibition, which is consistent with human cell line

data.

Supplementary Fig. 4 BRD2 KD enhances sensitivity to BET inhibition across murine

PDAC models. (A–H) Immunoblots validating BRD2 KD using independent shRNA constructs

(shBrd2 #1 and/or shBrd2 #2, as indicated) and dose–response curves for BETi showing reduced

cell viability in BRD2-depleted cells compared with shScr controls in mouse PDAC cell lines. **(I)**

Table of dose–response analyses showing LogIC$_{50}$ values with 95% confidence intervals (CI) and

ΔLogIC$_{50}$ values for BETi sensitivity following BRD2 KD across cell lines (ΔLogIC$_{50}$ calculated

relative to shScr within each cell line). **(J–O)** Representative colony formation assays and their

quantification (colony area and staining intensity) in mouse PDAC cell lines.

Supplementary Fig. 5 Body weight of mice was maintained during JQ1 treatment. (A) No

significant weight loss was observed in mice bearing tumors treated with vehicle or JQ1.

Supplementary Fig. 6 Basal BRD4/BRD2 levels and chromatin occupancy do not predict

JQ1 response. (A–B) Correlation analyses between basal BRD4 (A) or BRD2 (B) protein

expression levels and JQ1 sensitivity across human PDAC cell lines. No significant associations

were observed. **(C–D)** Correlation of *BRD4* and *BRD2* mRNA expression levels with JQ1

sensitivity across public cancer transcriptome datasets. Both factors showed weak or no

predictive value. **(E)** Summary table of LogIC$_{50}$ and 95% CI for JQ1, Molibresib, and Birabresib

in BETi-sensitive (mT3-2D) and BETi-resistant (mT23-2D) cells. **(F)** Venn diagram of BRD2 and

BRD4 ChIP-seq binding regions in TNBC (GSE131097, SUM149), highlighting substantial

overlap in line with our PDAC findings. **(G)** Metaprofiles of BRD4 (left) and BRD2 (right)

occupancy at BRD4-bound sites under DMSO or JQ1 conditions, suggesting BRD2 binding

persists at regions where BRD4 occupancy is displaced upon JQ1. **(H)** GOBP enrichment

analysis (ShinyGO) of the top 1,000 JQ1-responsive BRD2-bound regions annotated by GREAT,

revealing significant enrichment for RNA metabolic and biosynthetic processes.

Conclusions

Our study uncovered BRD2 upregulation as a conserved adaptive resistance mechanism to BET

inhibition across diverse cancer types. Through integrated transcriptomic and epigenomic

analyses, we demonstrated that BETi treatment induced BRD2 expression at the transcriptional

level, at least in part mediated by the NFYA transcription factor. BETi-mediated BRD2 induction

acts as a compensatory mechanism, diminishing the antitumor efficacy of BETi. Notably, BRD2

KD sensitizes cancer cells to BETi, suggesting the therapeutic potential of dual targeting

strategies that inhibit BRD2 alongside BETi to improve patient outcomes across multiple cancer

types.

Abbreviations

AML: acute myeloid leukemia

AT/RT: atypical teratoid/rhabdoid tumor

BC: breast cancer

BET: bromodomain and extra-terminal domain

BETi: BET inhibitors

BL: Burkitt lymphoma

CC: cervical cancer

CML: chronic myeloid leukemia

CRC: colorectal cancer

CRPC: castration-resistant prostate cancer

ES: Ewing sarcoma

ET: extra-terminal

GBM: glioblastoma

GIST: gastrointestinal stromal tumor

HCC: hepatocellular carcinoma

HNSCC: head and neck squamous cell carcinoma

KD: knockdown

Lum A: luminal A

MEL: melanoma

MM: multiple myeloma

NB: neuroblastoma

NFYA: Nuclear Transcription Factor Y subunit Alpha

NSCLC: non-small cell lung cancer

OC: ovarian cancer

PDAC: pancreatic ductal adenocarcinoma

RMS: rhabdomyosarcoma

SCLC: small cell lung cancer

SE: super-enhancers

TAD: topologically associating domain

T-ALL: T-cell acute lymphoblastic leukemia

TF: transcription factors

TNBC: triple-negative breast cancer

Declarations

Ethics approval and consent to participate

All animal studies were conducted in accordance with institutional and national guidelines and approved by the University of California, Davis (UC Davis) Institutional Animal Care and Use Committee (IACUC; Protocol #24016, approved August 22, 2024). UC Davis is accredited by the Association for Assessment and Accreditation of Laboratory Animal Care International (AAALAC) and maintains an Animal Welfare Assurance with the Office of Laboratory Animal Welfare (OLAW; Assurance Number D16-00272, A3433-01).

Consent for publication

All authors have reviewed and approved this manuscript and consent to its publication.

Availability of data and material

Raw and processed RNA-seq and ChIP-seq data generated in this study are available in GEO under accession numbers GSE307284 (RNA-seq) and GSE307285 (ChIP-seq).

Competing interests

The authors have no conflicts of interest to disclose.

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Authors' contributions

S.A. and C.I.H. conceptualized and designed the study. S.A., J.J., and E.L. performed the experiments and analyzed the data. S.A. drafted the original manuscript, and C.I.H. reviewed and edited the manuscript. All authors read and approved the final version of the manuscript.

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Fig. 1 Pan-cancer BRD2 upregulation upon BET inhibition.

A Pan-cancer *BRD2* expression in response to JQ1

B GSE78213
Differentially expressed genes
JQ1 vs. DMSO in castration resistant prostate cancer

C

Fig. 2 Pharmacological BET inhibition uniquely drives BRD2 upregulation in pancreatic cancer.

Fig. 3 NFYA as a candidate transcriptional regulator of BRD2 induction upon BET inhibition.

Fig. 4 BRD2 depletion sensitizes cancer cells to BET inhibition.

A

**Mouse
pancreatic cancer
mT23-2D**

shScr +
shBrd2 #1   +

BRD2

1.00   0.04

GAPDH

B

JQ1

Birabresib

Molibresib

ZEN-3694

CPI-0610

AZD5153

C

**Lung cancer
A549 (NSCLC)**

shScr +
shBRD2 #1   +

BRD2

1.00   0.50

GAPDH

D

**Brain cancer
U251 (GBM)**

shScr +
shBRD2 #1   +

BRD2

1.00   0.73

GAPDH

E

**Pediatric cancer
U2OS (OS)**

shScr +
shBRD2 #1   +

BRD2

1.00   0.63

GAPDH

F JQ1

G JQ1

H JQ1

I

**Human
pancreatic cancer
Capan-2**

shScr +
shBRD2 #1   +

BRD2

1.00   0.51

GAPDH

J

**Human
pancreatic cancer
CFPAC-1**

shScr +
shBRD2 #1   +

BRD2

1.00   0.26

GAPDH

K

**Human
pancreatic cancer
BxPC-3**

shScr +
shBRD2 #1   +

BRD2

1.00   0.12

GAPDH

L JQ1

M JQ1

N JQ1

Fig. 5 Genetic BRD2 perturbation enhances the antitumor effect of BET inhibition *in vivo*.

A

Mouse PDAC cells

Subcutaneous tumor injection

Left flank shScr   Right flank shBrd2

3 days

Control group
Vehicle
n = 10

Treatment group
JQ1 (50 mg/kg)
n = 10

Euthanize
Final tumor weight
Histologic examination

Schedule of treatment
(Intraperitoneal injection)

Schedule of tumor size measurement

Week   1   2   3

Treatment day
Rest day

B

shScr Vehicle

shScr JQ1

shBrd2 Vehicle

shBrd2 JQ1

C Tumor growth

Tumor volume (mm³)

- shScr Vehicle
- shScr JQ1
- shBrd2 Vehicle
- shBrd2 JQ1

Days after treatment

****

D Tumor weight

Tumor weight (g)

shScr Vehicle   shScr JQ1   shBrd2 Vehicle   shBrd2 JQ1

****

E

Vehicle   JQ1

shScr

shBRD2

F Ki67 staining quantification

Percentage of positive cells

shScr Vehicle   shScr JQ1   shBrd2 Vehicle   shBrd2 JQ1

****

Fig. 6 BRD2 retention sustains transcriptional programs in BETi-resistant pancreatic cancer cells.

Supplementary Fig. 1

Supplementary Fig. 2

Supplementary Fig. 3

A

	Gene	Function	hsa-u.4 / BRD2	hsa-r.6 / BRD2	hsa-m.7 / BRD2	hsa-m2.4 / BRD2
1	PPP1R10	protein phosphatase 1 regulatory subunit 10	8.9	8.1	5	7.7
2	NXF1	nuclear RNA export factor 1	8	6.9	2.9	7.1
3	POLR2A	RNA polymerase II subunit A	7.6	6.7	2.8	6.6
4	RBM22	RNA binding motif protein 22	7.5	7.8	2.6	5.3
5	FAM193A	family with sequence similarity 193 member A	7.4	7.4	1.1	5.6
6	BRPF3	bromodomain and PHD finger containing 3	7.1	5.5	1.7	6.9
7	PRRC2A	proline rich coiled-coil 2A	6.8	6.2	3.4	5.8
8	BTN2A1	butyrophilin subfamily 2 member A1	6.8	6.7	1.9	5.1
9	TRIM39	tripartite motif containing 39	6.7	5.9	2.8	5.9
10	SRCAP	Snf2 related CREBBP activator protein	6.6	5	1.8	6.6
11	BRPF1	bromodomain and PHD finger containing 1	6.6	6.1	1.9	5.4
12	RSRC2	arginine and serine rich coiled-coil 2	6.6	7.1	2.9	4.4
13	KDM2A	lysine demethylase 2A	6.5	6.5	2.7	4.9
14	SCAF4	SR-related CTD associated factor 4	6.2	5.8	1	5.1
15	FAM53C	family with sequence similarity 53 member C	6.2	5.4	2.1	5.6
16	CCNL1	cyclin L1	6.2	6.8	1.3	4.1
17	EP300	E1A binding protein p300	6.2	5.4	1.9	5.5
18	ABCF1	ATP binding cassette subfamily F member 1	6.2	6	2.2	4.8
19	PHF12	PHD finger protein 12	6.1	5.8	1.2	5
20	CNOT4	CCR4-NOT transcription complex subunit 4	6.1	7.5	1.5	3.2
21	TRIM26	tripartite motif containing 26	6.1	7	2.6	3.7
22	ZNF592	zinc finger protein 592	6	4.1	0.6	6.4
23	ATXN2L	ataxin 2 like	5.9	5.1	1.9	5.3
24	ZNF335	zinc finger protein 335	5.8	5.3	1.8	5
25	SETD5	SET domain containing 5	5.8	6	2.3	4.2
26	CDK12	cyclin dependent kinase 12	5.8	6	0.6	4.2
27	NUP153	nucleoporin 153	5.8	5.6	1.8	4.5
28	CREBBP	CREB binding protein	5.8	5.2	0.9	5
29	SRRM2	serine/arginine repetitive matrix 2	5.8	4	5.1	6.2
30	NFYA	nuclear transcription factor Y subunit alpha	5.7	5.6	1.7	4.4

B correlation of expression pattern

coex z = 5.62

BRD2 / NFYA

C *Nfya* — Normalized counts — DMSO / JQ1 — ***

D Chr6:32,955,022-33,021,623 (hg38)

HEK293T — H3K4me3 (GSM4445878) [0 - 114]
HEK293T — RNA Pol II (GSM4602027) [0 - 271]
K562 — NFYA (GSM935433) [0 - 85]
GM12878 — NFYA (GSM935506) [0 - 28]
HeLa-S — NFYA (GSM935508) [0 - 160]
BRD2

NFYA ChIP-qPCR

E mT5-2D — shScr / shNfya — DMSO JQ1 DMSO JQ1 — NFYA, BRD2, GAPDH

F mT23-2D — shScr / shNfya — DMSO JQ1 DMSO JQ1 — NFYA, BRD2, GAPDH

Supplementary Fig. 4

Supplementary Fig. 4

I

LogIC$_{50}$ and ΔLogIC$_{50}$ of BETi sensitivity after BRD2 KD across cell lines

Drugs	Cell lines		LogIC$_{50}$ (M)	95% CI	ΔLogIC$_{50}$
JQ1	mT23-2D	shScr	-4.556	-4.733 to -4.310	
		shBrd2 #1	-5.634	-5.760 to -5.506	1.078
	mT3-2D	shScr	-5.455	-5.769 to -5.104	
		shBrd2 #1	-6.258	-6.464 to -6.040	0.803
	mT4-2D	shScr	-4.667	-4.805 to -4.534	
		shBrd2 #1	-6.081	-6.291 to -5.864	1.414
		shBrd2 #2	-5.956	-6.128 to -5.782	1.289
	mM1-2D	shScr	-4.896	-5.018 to -4.776	
		shBrd2 #1	-5.704	-5.828 to -5.576	0.808
		shBrd2 #2	-5.112	-5.287 to -4.920	0.216
	mM6-2D	shScr	-4.486	-4.565 to -4.410	
		shBrd2 #1	-4.891	-5.047 to -4.718	0.405
		shBrd2 #2	-4.618	-4.844 to -4.305	0.132
	A549	shScr	-4.092	NE	
		shBRD2 #1	-5.105	-5.317 to -4.887	1.013
	U251	shScr	-3.979	NE	
		shBRD2 #1	-5.305	-5.576 to -5.022	1.326
	U2OS	shScr	-4.382	-4.538 to -4.249	
		shBRD2 #1	-6.163	-6.493 to -5.842	1.781
	Capan-2	shScr	-3.13	-3.314 to -2.871	
		shBRD2 #1	-4.845	-5.080 to -4.612	1.715
	CFPAC-1	shScr	-4.367	-4.514 to -4.233	
		shBRD2 #1	-4.839	-4.528 to -4.229	0.472
	BxPC-3	shScr	-5.386	-5.603 to -5.174	
		shBRD2 #1	-6.425	-6.638 to -6.224	1.039
Birabresib	mT23-2D	shScr	-4.551	-4.744 to -4.362	
		shBrd2 #1	-5.127	-5.316 to -4.933	0.576
	mT4-2D	shScr	-4.701	-4.845 to -4.561	
		shBrd2 #1	-5.949	-6.135 to -5.758	1.248
		shBrd2 #2	-5.825	-5.995 to -5.651	1.124
Molibresib	mT23-2D	shScr	-3.792	-4.045 to -3.236	
		shBrd2 #1	-4.311	-4.546 to -4.013	0.519
ZEN-3694	mT23-2D	shScr	-4.104	-4.401 to -3.686	
		shBrd2 #1	-4.52	-4.742 to -4.271	0.416
CPI-0610	mT23-2D	shScr	-4.003	-4.326 to -3.412	
		shBrd2 #1	-4.966	-5.265 to -4.619	0.963
	mT4-2D	shScr	-3.467	-3.872 to -2.656	
		shBrd2 #1	-5.651	-6.416 to -4.673	2.184
		shBrd2 #2	-5.787	-6.104 to -5.438	2.32
AZD5153	mT23-2D	shScr	-4.481	-4.708 to -4.229	
		shBrd2 #1	-5.497	-5.732 to -5.250	1.016
	mT4-2D	shScr	-4.832	-4.983 to -4.678	
		shBrd2 #1	-6.425	-6.709 to -6.124	1.593
		shBrd2 #2	-6.342	-6.539 to -6.141	1.51

IC$_{50}$: Half maximal inhibitory concentration
CI: Confidence interval
NE: Not estimable

Supplementary Fig. 4

Supplementary Fig. 5

A

Supplementary Fig. 6

A BRD4 basal level vs JQ1 response

$r = -0.1775$
p value = 0.649

B BRD2 basal level vs JQ1 response

$r = -0.2793$
p value = 0.4668

E

Drugs	Cell lines	LogIC$_{50}$ (M)	95% CI
JQ1	mT3-2D	-6.989	-7.139 to -6.841
	mT23-2D	-4.583	-4.775 to -4.419
Molibresib	mT3-2D	-5.875	-5.969 to -5.779
	mT23-2D	-3.605	-3.754 to -3.393
Birabresib	mT3-2D	-6.598	-6.713 to -6.479
	mT23-2D	-4.623	-4.751 to -4.496

C $r = -0.102$
p-value = 0.043

D $r = -0.058$
p-value = 0.254

F GSE131097 Breast cancer SUM149

BRD2 and BRD4 shared regions 9330

BRD4 binding regions 3579

BRD2 Binding regions 30804

G BRD4 occupancy
BRD4 ChIP on BRD4 peaks at BRD4 binding sites

SUM149 DMSO
SUM149 JQ1

BRD2 occupancy
BRD2 ChIP on BRD4 peaks at BRD4 binding sites

SUM149 DMSO
SUM149 JQ1

H ShinyGO analysis: GO Biological Process
Top 5% BRD2 loss sites after JQ1 (mT3-2D)