BET inhibitors repress expression of interferon-stimulated genes and synergize with HDAC inhibitors in glioblastoma

Olga Gusyatiner, Pierre Bady, Minh D. T. Pham, Yvonne Lei, Jungyeon Park, Roy T. Daniel, Mauro Delorenzi, and Monika E. Hegi

Neuroscience Research Centre, Lausanne University Hospital and University of Lausanne, Lausanne, Switzerland (O.G., P.B., M.D.T.P., Y.L., J.P.); Service of Neurosurgery, Lausanne University Hospital and University of Lausanne, Lausanne, Switzerland (O.G., P.B., R.T.D., M.E.H.); Bioinformatics Core Facility, SIB Swiss Institute of Bioinformatics, Lausanne, Switzerland (P.B., M.D.); Department of Oncology, University of Lausanne, Lausanne, Switzerland (M.D.); Swiss Cancer Center Léman (SCCL), Lausanne, Switzerland (O.G., P.B., M.D., M.E.H.)

Corresponding Author: Monika E. Hegi, PhD, Neuroscience Research Center and Service of Neurosurgery, CHUV, chemin des Boveresses 155, 1066 Epalinges, Switzerland (monika.hegi@chuv.ch).

Abstract

Background. The development of rational combination therapies is key to overcome inherent treatment resistance of glioblastoma (GBM). We aim at identifying new druggable targets by disturbing GBM cells with inhibitors of bromodomain and extra-terminal motif (BET) proteins to reveal cancer-relevant vulnerabilities that may sensitize to a second drug. BET proteins are epigenetic modulators and have been associated with proto-oncogene overexpression in cancer.

Methods. A GBM-derived sphere-line was treated with the BET inhibitor (BETi) JQ1 over a time-course of 48 hours, followed by RNA-sequencing. Four chromatin marks were investigated by chromatin immunoprecipitation followed by sequencing (ChIP-seq). Signatures of interest were functionally validated in vitro and in orthotopic xenografts. Combination therapies were evaluated for synergistic effects.

Results. Cancer-relevant pathways significantly modulated by JQ1 comprised interferon alpha (IFN-α) response genes and response signatures to histone deacetylase inhibitors (HDACi). The IFN-signature was reminiscent of a GBM-derived IFN-signature comprising CD274 (PD-L1). Functional pathway analysis suggested that JQ1 was acting directly on the transcriptional level of IFN-response genes and not via the canonical JAK/STAT pathway. This was in line with JQ1 modulated expression and BRD4 and Pol II occupancy at IFN-signature genes, supporting a direct mechanistic interaction. Finally, we showed that combining HDACi with JQ1 acts synergistically in reducing cell viability of GS-lines.

Conclusions. Our approach identified BETi-induced vulnerabilities in cancer-relevant pathways, potentially amenable to synergistic combinatorial therapy, such as combination with HDACi. The direct inhibitory effect of BETi on IFN-responsive genes in GBM cells, including CD274, indicates modulation of the tumor immune landscape and warrants further studies.

Key Points

- BRD4 regulates IFN-stimulated gene signature.
- Combination of BETi and HDACi is synergistic.
Importance of the Study

Glioblastoma (GBM) is one of the most difficult tumors to treat and is notorious for its resistance to treatment, despite recent insights on the evolution of the disease gained from multidimensional omics. The failures of single agents tested in clinical trials have dramatically shown that combination therapies are necessary to improve the outcome for patients affected with these treatment-resistant tumors. However, the rational choice for successful combination therapy is the big challenge. Here, we present our approach to harness epigenetic changes that may be targeted by a new class of drugs, bromodomain and extra-terminal (BET) inhibitors. BET proteins are epigenetic regulators that have been associated with overexpression of proto-oncogenes in a variety of cancers and seem to be promising cancer-relevant targets. We disturb the system with a small molecule BET inhibitor, which elucidates emerging pathway vulnerabilities that may become targetable with a second drug.

Methods

Cell Culture

Human GBM-derived sphere (GS)-lines LN-2207GS, LN-2683GS, LN-2540GS, LN-2669GS, LN-3704GS, LN-3708GS, and adherent GBM cell lines LN-18 and LN-428 were established and molecularly characterized in our laboratory, with ethics approval (CER-VD, PB_2017-00240/F-25/99). U87MG was obtained from ATCC and BS-153 from Adrian Merlo. All lines were authenticated as described and tested mycoplasma-free (MycopAlert Kit, Lonza). Adherent lines were grown in Dulbecco’s modified Eagle's medium (DMEM, Gibco), 5% fetal calf serum (Hyclone, Thermo Fisher Scientific). Experiments with GS-lines were performed under stem cell conditions in DMEM/F12 (Invitrogen) supplemented with B27 (Gibco) and trichostatin A (TSA, Sigma). For adherent GBM cell lines (+)-JQ1, or (−)-JQ1 was added starting with 30 µM, and 5 mM, respectively.

Cell Viability Assays

Lethal concentrations of actinomycin D (1 µg/mL) or DMSO were used as 0% and 100% viability controls. Cell viability was measured after 5 days using the CellTiter-Blue (Promega). Lethal concentrations of actinomycin D (1 µg/mL) or DMSO were used as 0% and 100% viability controls. Experiments with GS-lines were performed under stem cell conditions in DMEM/F12 (Invitrogen) supplemented with B27 (Gibco) (1/50) and growth factors as previously described.

Spheres were separated into single-cell suspension using Accutase (Gibco). GS-lines were primed with 1000 units/mL interferon alpha (IFN-α) (Peprotech) or 20 ng/mL IFN-γ (GenWay Biotech). (+)-JQ1, its inactive enantiomer (−)-JQ1 (ApexBio Technology LLC) and trichostatin A (TSA, Sigma) were dissolved in dimethyl sulfoxide (DMSO) at 10 mM and 5 mM, respectively.

Cell Assays

The senescence-associated β-galactosidase assay on adherent GBM cells was performed according to the manufacturer’s instructions (BioVision). Immunofluorescence of sphere lines used anti-TUJ1 (Cell Signalling, 1:400) and fluorechrome-conjugated anti-rabbit Alexa647 (Abcam, 1:500). The BrdU incorporation assay was performed using FITC (fluorescein isothiocyanate)-labeled BrdU antibody and 7-AAD (7-aminomethylcoumarin D, Pharmingen™ BrdU Flow Kit, BD Biosciences) and was analyzed by fluorescence-activated cell sorting (FACS, Calibur). Experimental details are available in Supplementary Methods and Table S1.
cell sorting (FACS, Calibur). Experimental details are available in the Supplementary Methods.

Cell Viability Assays

For adherent GBM cell lines (+)-JQ1, or (−)-JQ1 was added in a 10-point serial dilution of 1:3, starting from 30 µM. GS-lines were treated using a 10-point serial 1:2 dilution, starting with 30 µM of (+)-JQ1 or (−)-JQ1 (see details in Supplementary Methods). After 72 hours, cell viability was measured using the CellTiter-Blue or CellTiter-Glo Assay (Promega). Lethal concentrations of actinomycin D (1 µg/mL) or DMSO were used as 0% and 100% viability controls. A curve-fitting algorithm with variable Hill slope was applied to derive IC_{so} values. Experiments were repeated 3 times.

Drug Combination Assay

GS cells were treated using a 7-point serial 1:2 dilution starting with 20 µM, employing a chessboard pattern (8 x 8), and every concentration of JQ1 was tested with every concentration of TSA. DMSO served as control. Cell viability was measured after 5 days using the CellTiter-Blue Assay. For the 10-day assay, JQ1 and TSA were added at a serial dilution starting from 2 µM. After 5 days, drugs were renewed as detailed in the Supplementary Methods. The combination score^{22,23} was determined using the R package synergyfinder (combination of Lowe and Bliss combination indexes)^24. Experiments were repeated 3 times.

Neurosphere Formation Assay

Adapting a protocol by Cheng et al.,^{13} single cells were seeded into a 48-well plate in 100 µL F12/B27 medium (containing EGF [epidermal growth factor]/FGF [fibroblast growth factor]) with 100 (LN-2207GS, LN-2540GS, LN-2683GS) or 1000 cells/well (LN-2669GS). JQ1, diluted in 100 µL F12/B27 to a final concentration of 0, 0.25, 0.5, 1, 2, or 4 µM, was added, with 4 replicates per condition. EGF/FGF were replenished on days 8 and 15 (20 ng/mL). The spheres with a diameter ≥50 µm were counted after 14 and 21 days. The experiment was repeated 3 times.

RNA Extraction, qRT-PCR

Total RNA isolation and qRT-PCR were performed as described previously^{21} using primers compiled in Supplementary Table S1. The expression levels were normalized to GAPDH and to the geometrical mean of hGAPDH and C1orf43 for cDNA derived from xenografts.

Western Blot and Nuclear Fractionation

Cells were collected by centrifugation for GS-lines and by scraping for adherent cells. Westerns were done as described^{21,25} and probed with respective antibodies (Supplementary Methods). Nuclear and cytoplasmic fractionation was performed using the NE-PER Kit (Thermo Scientific) with modifications detailed in the Supplementary Methods.

RNA Sequencing and Differential Expression Analysis

LN-2683GS cells were treated with (+)-JQ1 (1 µM) or DMSO and harvested at 0, 4, 12, 24, and 48 hours. RNA-sequencing was accomplished at the Genomic Technology Facilities (University of Lausanne). Data were preprocessed following standard pipeline and recommendations from bbio-nextgen (version 0.8, http://bbio-nextgen.readthedocs.org/en/latest/). Differential expression analysis used a model with full interaction between treatment and time (edgeR package, version 3.12.0). Genes with Bonferroni adjusted P values <.001 were considered differentially expressed. Gene-set enrichment analysis (GSEA) was performed (PGSEA package, version 1.44.0; GAGE package, version 2.20.1) using The Molecular Signatures Database (MSigDB v5, Broad Institute). Gene sets with FDR (false discovery rate)-adjusted P values <.05 were considered significant. Due to patient privacy protection, the raw sequencing data will be made available upon request. See the Supplementary Methods for details.

Chromatin Immunoprecipitation (ChIP)

LN-2683GS cells were treated with 1 µM JQ1, DMSO, and 1000 units/mL IFN-α for 2 hours. U87MG were treated with 1 µM JQ1 or DMSO for 24 hours. Cells were cross-linked and CHiP was performed using iDeal ChIP-seq kit for Transcription Factors (Diagenode) as detailed in the Supplementary Methods, using the following antibodies: anti-Pol II (Cell Signaling, D8L4Y, 1:50), anti-BRD4 (Bethyl Laboratories, A301, 1:60), anti-H3K4me3 (Diagenode, C15410003, 1:300), anti-H3K27ac (Abcam, ab4729, 1:300 or 1:600), control IgG (1:600, Diagenode).

ChIP Sequencing Data Analysis

The preprocessing of ChIP-seq data was performed following the pipeline suggested by project bbio-nextgen (version 1.1.6, http://bbio-nextgen.readthedocs.org/). The binding affinity matrix was built by peaks identified in at least 2 samples. The raw read counts were normalized and batch corrected to process ChIP-seq data prior to further analysis. The global effect of JQ1 treatment, IFN stimulation, and their interaction with the IFN-signature for the 4 chromatin marks was tested using Permutational Multivariate Analysis of Variance based on Euclidean distance matrices (ADONIS with 999 permutations^{26}). Corresponding RNA-seq data, normalized and batch corrected was used in the heatmap and to compute pairwise Spearman correlation between Pol II marks and expression of the related genes. The ChIP-seq normalized datasets were weighted by their total inertia for the simultaneous heatmap representation as used in MFA (multiple factor analysis)^27. See details in the Supplementary Methods.
Fig. 1 Effects of JQ1 on glioma sphere lines. (A) Cell viability of LN-2207GS, LN-2540GS, and LN-2683GS was determined after 72 hours’ incubation with 3-fold serial dilutions of (+)-JQ1 and (−)-JQ1, normalized to DMSO control. Data represent a mean of 3 independent experiments; error bars are SEM. (B) Quantification of neurosphere formation assay of LN-2207GS, LN-2540GS, LN-2683GS, and LN-2669GS. The spheres >50 µm in diameter were counted 21 days after (+)-JQ1 addition. Data represent mean of 4 wells; error bars are SD; representative of at least 3 independent experiments. P values, 2-way ANOVA, adjusted for multiple testing, statistical hypothesis Dunnett test. (C) Western blot for total PARP, cleaved PARP (c-PARP), and tubulin (loading control) of LN-2683GS and LN-2207GS, 10 days after treatment with 1 µM (+)-JQ1 or DMSO. (D) Representative image of neurospheres taken after 21-day incubation of LN-2683GS with 0.5 µM (+)-JQ1 and DMSO. (E) BrdU incorporation assay
Orthotopic Xenografts

Orthotopic xenografts were obtained with $10^5$ U87MG cells transplanted into the striatum of 6-NOD-SCID-gamma knockout mice as previously described. Mice were injected i.p., with JQ1 (100 mg/kg) 4 hours before sacrifice. The protocol was approved by the local authorities (VD-3266). See the Supplementary Methods for details.

Statistical Analysis

Statistical analysis was performed using GraphPad Prism 7 Software and R-3.2.2. Bioconductor packages for specific tasks are listed in the relevant sections. The Student t test was used to compare the variables between 2 groups, and 2-way ANOVA for more complex experimental design. $P$ values <.05 were considered statistically significant: significance is indicated with asterisks: *$P$ < .05, **$P$ < .01, ***$P$ < .001, and ****$P$ < .0001. Data are presented as mean values, error bars represent standard deviation unless indicated otherwise. Similarity between distance matrices was evaluated by RV (vectorial correlation) coefficient (values between 0 and 1), using R package ade4 for pairwise RV coefficient permutation tests.

Results

JQ1 Treatment Reduces Sphere Formation and Viability of GBM Cells In Vitro

First, we evaluated the effects of JQ1 on cell viability of GS-lines (Figure 1A). They appeared more resistant to JQ1 at 72 hours as compared to adherent GBM cell lines, with IC50 ranging between 5.8-18.6 µM and 1.4-11.6 µM, respectively (Supplementary Table S2). A proportion of cells remained viable at high JQ1 concentrations that were assessed otherwise. Similarity between distance matrices was evaluated by RV (vectorial correlation) coefficient (values between 0 and 1), using R package ade4 for pairwise RV coefficient permutation tests.

BET Protein Inhibition Causes Extensive Changes in the Transcriptome

To gain insights into biological changes we performed differential gene expression profiling by RNA-sequencing of LN-2683GS treated with 1 µM JQ1 over a time course of 4, 12, 24, and 48 hours (Figure 2A). This yielded 2709 differentially expressed genes (1278 down, 598 up) using an additive model with interaction between time and treatment (log-likelihood ratio test, Bonferroni, $P$ < .001; fold change >2 for at least 1 time point) (Supplementary Table S3). Pathway analysis of differentially expressed genes revealed significant modulation of gene sets, such as those associated with particular epigenetic features of neural progenitor cells (NPCs) and embryonic stem (ES) cells, gene sets modulated by epigenetic interference, and inflammation-related signatures (Figure 2C). In JQ1-treated LN-2683GS cells, these comprised downregulation of interferon-stimulated genes (ISG) enriched for Hallmark interferon alpha response (HIAR) signature genes (Figure 2D). Furthermore, we detected enrichment in gene sets associated with cellular response to HDACi (Figure 2C).

Interestingly, we found that the JQ1-modulated HIAR gene set overlapped largely (25 genes) with our previously reported ISGs expression signature (G12) identified in a human GBM dataset (Figure 2D). This GBM-derived IFN-signature (G12) shows high similarity of co-expression among publically available GBM datasets (RV values > 0.8; $P$ value = .001; Supplementary Figure S1A and B) and comprises in addition gene sets associated with IFN-γ response, such as CD274, encoding PD-L1 (programmed death-ligand 1), known for its importance in tumor immune evasion.

Expression of ISG Signature Is Repressed Upon JQ1 Treatment

To validate the downregulation of ISGs upon JQ1 treatment, we boosted IFN signaling by priming GS-lines with orthotopic GBM cells.
Fig. 2 Transcriptome analysis of JQ1 effects. (A) Experimental setup of differential gene expression analysis by RNA-seq. LN-2683GS cells were treated with 1 µM (+)-JQ1 or DMSO for 4, 12, 24, and 48 hours, the experiment was repeated 3 times, followed by RNA-seq. (B) Principal coordinate analysis of the raw read counts of RNA-seq data. PC1, PC2: principal coordinates 1 and 2. (C) Top 15 downregulated (blue) and upregulated (red) gene sets based on GSEA (q-value, Bonferroni corrected for multiple testing). Gene sets in bold were selected for further analysis. (D) Heatmap of log2FC (FC: fold change) of expression of IFN-α response gene set (HIAR). Genes marked in red overlap with our GBM IFN-signature (G12, Supplementary Figure S3). Abbreviations: DMSO, dimethyl sulfoxide; GBM, glioblastoma; GSEA, gene-set enrichment analysis; HIAR, hallmark interferon alpha response; IFN-α, interferon alpha.
Fig. 3  Inhibition of interferon-stimulated genes upon JQ1 treatment in vitro and in vivo. Expression of CD274, MX1, and OAS1 (qRT-PCR) in LN-2683GS and LN3704GS primed 4 hours with IFN-α, followed by 4 hours' JQ1 (1 µM) treatment (A). Mean of 3 independent experiments, SD; 2-way ANOVA, multiple testing correction, Sidak test. (B) Western blot for PD-L1 and GAPDH for LN-2683GS and LN-3704GS treated as in A, representative of 3 independent experiments. Arrows indicate specific band for PD-L1. (C) Expression of endogenous ISGs, CD274, MX1, and OAS1 (qRT-PCR) in adherent GBM cell lines U87MG and LN-18 after 48 hours' treatment with 1 µM JQ1 (C) (2-way ANOVA, multiple testing correction, Sidak test). (D) Western blot for PD-L1 in U87MG and LN-18 treated as in (C). (E) U87MG derived orthotopic xenograft immunostained for PD-L1.
IFN-α for 4 hours, followed by 4 hours of JQ1 exposure. We chose OAS1 and MX1 as markers of IFN-response (Figure 3A), both carry an interferon-response element (ISRE) and also belong to the G12 signature (Supplementary Figure S3A). Furthermore, we included CD274 (PD-L1) in the analysis for its importance in anti-tumor immunity. We detected a strong reduction of expression of the ISGs OAS1, MX1, and CD274 in JQ1-treated GS-lines as compared to DMSO (Figure 3A). Similar results were obtained with IFN-γ priming, confirming repression of CD274 by JQ1 treatment in the GS-lines on the RNA and protein level (Figure 3B).

To exclude possible effects of JQ1 on JAK/STAT signaling per se, we examined pSTAT1 levels in the nucleus of GS-lines after priming with IFN-α or IFN-γ. While total STAT1 and pSTAT1 were increased in the nucleus after IFN-priming, we did not observe a change upon JQ1 treatment in the GS-lines as compared to DMSO (Supplementary Figure S4).

IFN-α Stimulation Enhances BRD4 Occupancy at ISGs

To evaluate whether the ISGs are regulated through BRD4-mediated transcription, we performed ChIP-seq for BRD4 in LN-2683GS cells, with or without 2 hours’ treatment with IFN-α, or JQ1, or their combination. Modulation of BRD4 coverage was observed at our marker genes CD274, MX1, and OAS1 under the 4 experimental conditions, with increased binding upon IFN-α stimulation, and decrease under JQ1 treatment. Acetylation of H3K27 was increased by IFN-α stimulation, along with Pol II recruitment at the active promoters of target genes, while JQ1 treatment did not seem to alter H3K27ac levels, but reduced Pol II occupancy at the active promoters defined by H3K4me3 peaks. The tracks of the consensus peaks for BRD4, Pol II, and the histone marks H3K27ac and H3K4me3 are illustrated for the area within 20 kb of the transcription start sites (TSSs) (Supplementary Figure S5). Next, we explored whether the whole ISG gene signature was modulated by BRD4-binding under the 4 experimental conditions and performed GSEA of BRD4 occupancy using consensus peaks. The results from genome-wide differential binding analysis were significant for the expression of the ISG signature (additive effect), for treatment with JQ1 (adj. P value = .009), for priming with IFN-α (adj. P value = .0018), and interaction between JQ1 treatment and IFN-α stimulation (adj. P value = .0021). In order to illustrate that BRD4 occupancy is similarly modulated across the IFN-signature genes under the 4 experimental conditions, a heatmap of the BRD4 consensus peaks associated with these ISG signature genes is shown, complemented with respective heatmaps for RNA-seq and ChIP-seq for Pol II, H3K27ac, and H3K4me3 (Figure 4). Similar to the 3 marker genes, CD274, MX1, and OAS1, most of the IFN-signature genes revealed increased BRD4 occupancy with IFN-α stimulation that was repressed upon JQ1 treatment, which was also reflected on the respective RNA expression levels. Furthermore, IFN-α stimulation triggered acetylation of H3K27 along with Pol II binding, whereas JQ1 treatment did not significantly alter H3K27ac levels, but significantly reduced Pol II occupancy. The respective multivariate analysis of the effects on the IFN-signature by experimental modality is summarized in Supplementary Table S4 for the 4 chromatin marks and RNA expression.

Endogenous Expression of ISGs Is Repressed Upon JQ1 Treatment In Vitro and In Vivo

To investigate whether JQ1 represses endogenous expression of ISGs, we treated U87MG and LN-18 cell lines, which express CD274 (PD-L1) endogenously, with 1 μM JQ1 for 48 hours. As a result, CD274, OAS1, and MX1 expression were largely reduced, and PD-L1 protein was diminished (Figure 3C and D). In accordance, ChIP-seq analysis of U87MG, visualized for CD274 (Supplementary Figure S6), indicates binding of BRD4, H3K27ac, Pol II, and H3K4me3 at the TSS before JQ1 treatment (IFN-untreated).

To determine the effects of JQ1 on ISG expression in vivo, we administered JQ1 (100 mg/kg, i.p.) or vehicle control to mice bearing orthotopic U87MG-derived xenografts for 4 hours before their sacrifice. Using human-specific primers, a significant decrease in CD274 and OAS1 expression was detected in the tumor xenografts upon JQ1 treatment, whereas no significant difference was observed for MX1 and c-Myc (Figure 3F). Together with the strong JQ1-induced upregulation (4-fold) of HEXIM1, a pharmacodynamic marker for BETi exposure, the data suggested that a pharmacologically relevant concentration of JQ1 was reached in the orthotopic tumor xenografts.

JQ1 and TSA Act Synergistically to Reduce Cell Viability of GS-Lines

The JQ1-induced modulation of gene sets related to HDACi-associated cellular response signatures motivated...
us to test whether JQ1 sensitizes GS-lines to HDACi treatment. We therefore tested TSA, a small molecule pan-HDAC inhibitor in combination with JQ1 in the GS-lines LN-2683GS and LN-3708GS. A scheme of 64 possible combinations of JQ1 and TSA was employed, ranging from 0 \mu M to 2 \mu M for a 10-day assay (Figure 5A–C, Supplementary Figure S7). The dose-response matrix (Figure 5B) was analyzed to determine the range of concentrations with synergistic drug interactions (Figure 5C). Based on the performed analysis, synergistic drug interactions explained 5% of cell deaths in LN-2683GS after 3 days, and 9% after 10 days, and for LN-3708GS 15% and 4% after 3 and 10 days, respectively (Supplementary Figure S7).

Discussion

BETis have demonstrated efficacy in NUT-midline carcinoma that is driven by the oncogenic BRD4–NUTM1 translocation. BETis are currently studied in other solid tumors including glioma, where they are tested as single agents or in combination with standard radio-chemotherapy.
Fig. 5 Combination of JQ1 with TSA shows synergy in reducing cell viability of GS-lines. (A) Single-drug cell viability curves for LN-2683GS treated with serial 2-fold dilutions of TSA or JQ1, respectively, starting with 2 µM for 10 days. Data represent mean of 3 plates (technical replicates). (B) Combination matrix of cell inhibition by JQ1 and TSA. 100% is inhibition of LN-2683GS cells treated with the lethal dose of 1 µg/mL actinomycin D, 0% is inhibition of LN-2683GS exposed to DMSO. Data represent mean of 3 wells (technical replicates). (C) Combination landscape, synergistic range of TSA and JQ1 concentrations in red, additive range in green, representative of 3 independent experiments. The synergy score is estimated by the zero interaction potency (ZIP) model, where the expected response corresponds to an additive effect as if the 2 drugs do not affect the potency of each other. The ZIP score has a unit of percentage inhibition and can be interpreted as the proportion of cellular responses that can be attributed to the drug interactions; ZIP score = 0 no interaction, ZIP score >0 synergistic, and <0 for antagonistic effect. Abbreviations: DMSO, dimethyl sulfoxide; GS, glioblastoma-derived sphere; TSA, trichostatin A.
Hence novel strategies for rational combination strategies are warranted.41,42

In GBM models, the BETis JQ1, I-BET151, OTX015, and the BET-degrader, dBET6, have displayed some positive preclinical efficacy.13–14,43,44 Our data support the anti-proliferative activities of JQ1 in both adherent cell lines and patient-derived GS-lines, showed apoptotic cell death in patient-derived GS-lines, and uncovered a differentiation-like phenotype in some GS-lines.

In this study, we focus on predicting rational combination strategies by identifying pathway vulnerabilities induced by the BETi JQ1. Pathway analysis of genes disturbed by JQ1 treatment uncovered an IFN-α-response signature among the top downregulated gene sets, suggesting that BETi suppresses transcription of a significant proportion of IFN-response genes in GBM. This is of interest, as around 50% of GBM display enhanced expression of a reminiscent IFN-signature, as we have reported previously (Supplementary Figure S3).35 Indeed, based on the integrated analysis of ChIP-seq and RNA-seq data, a large fraction of IFN-response genes are activated through acetylation of H3K27 and BRD4 recruitment upon IFN stimulation. Thus, transcription of ISGs, including CD274, is directly repressed by JQ1-induced release of BRD4 from the chromatin. Of note IFN-α-induced acetylation of H3K27 at IFN-signature genes was not significantly affected by JQ1 treatment, in accordance with a recent report suggesting that BETi blunts transcription, but retains enhancer-promoter contact.46 JQ1 impaired expression of endogenous, as well as interferon-induced ISGs. This suggests that BRD4 chromatin occupancy of ISGs is a key regulator of response to signals from the GBM microenvironment, and may regulate intrinsic ISGs expression. Previously, we have shown that PD-L1 is detectable by immunohistochemistry in the majority of human GBMs, while higher CD274 expression is enriched in the mesenchymal GBM subtype.46 Of note, the mesenchymal GBM subtype has a more prominent component of brain-resident microglia and tumor-infiltrating immune cells.47 Interestingly, Hogg et al48 reported a stronger tumor response to BETi in immune-competent tumor-bearing mice as compared to immunocompromised mice, which incited the authors to combine BETi and immune checkpoint inhibition targeting PD-1 that further improved outcome. Along similar lines BETi has been suggested to modulate T cells, thereby improving the effectiveness of checkpoint inhibitors in a melanoma model.49 On the other hand, conservation of an IFN-γ signature in melanoma was associated with better response to immune checkpoint blocking therapy in patient cohorts.50 Hence, the effects of BETis on immune cells in the GBM microenvironment will need to be addressed, particularly as immune checkpoint inhibiting antibodies are currently tested in multiple GBM trials (NCT02617589; NCT02667587).

In our study, a data-driven approach based on differential gene expression analysis was used to predict a therapeutic potential of combining BETi and HDACi. In fact, a synergy between HDACi and BETi was previously reported from models of several tumor types.33,51 In our experiments, the combination of JQ1 and the pan-HDAC inhibitor TSA showed synergism in reducing cell viability. In line, a combination of BETi and HDACi treatment has been reported to modulate the expression of pro- and anti-apoptotic genes as compared to single-agent treatments.33 Of note, this study demonstrated inefficacy of the combination of the HDACi Panobinostat and BETi OTX015, when compared to OTX015 alone in orthotopic GS-line-derived xenografts, while the same combination reduced subcutaneous tumor growth more efficiently than the single agents, indicating limited penetrance of the BBB.33 The poor BBB penetration of HDACi may explain in part why the pan-HDACi Vorinostat, FDA approved for the treatment of lymphoma, failed in trials for recurrent GBM patients as a monotherapy, or in combinations in newly diagnosed GBM.52,53 Collectively, our study defines BETi target genes and provides evidence that BRD4 is involved in the transcriptional regulation of a GBM interferon response signature. We have also shown that the combination of an HDACi with a BETi, in therapeutically relevant concentrations, can synergistically reduce the viability of GS-lines. However, the prospective translation to the clinic may only be possible once both BETi and HDACi are optimized for penetrance of the BBB.

Given that epigenetic drugs modulate the inherent, individual expression profile of tumors (and the tumor environment), specific combination therapies will require respective patient selection based on biomarker signatures. This has also been suggested by the authors of a phase II trial for HDACi in newly diagnosed GBM. Although their study was negative, a post hoc subgroup analysis searching expression signatures as predictive factors identified subgroups of patients that may have benefitted from treatment.42,52

Supplementary Material

Supplementary material is available at Neuro-Oncology online.

Keywords

BET inhibitors | epigenetics | glioblastoma | interferon-stimulated genes | synergistic combination therapy

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Authorship statement. O.G. and M.E.H. designed the study. O.G. performed the experiments, analyzed and interpreted the data, and wrote the manuscript. M.D.T.P., Y.L., and J.P. performed the experiments and analyzed the data. P.B. performed data analysis (RNA-seq, ChIP-seq) interpreted the bio-statistical analyses. R.T.D. provided study material. M.D. interpreted the bio-statistical analysis. M.E.H. directed the study, interpreted the data, and wrote the manuscript. All authors contributed to manuscript writing.

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