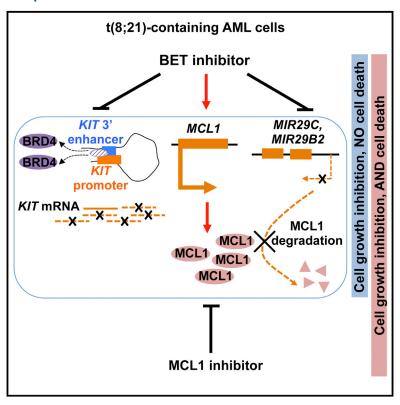
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High-Resolution Mapping of RNA Polymerases Identifies Mechanisms of Sensitivity and Resistance to BET Inhibitors in t(8;21) AML

Graphical Abstract



Authors

Yue Zhao, Qi Liu, Pankaj Acharya, ..., Ming-Ming Zhou, John T. Lis, Scott W. Hiebert

Correspondence

scott.hiebert@vanderbilt.edu

In Brief

Zhao et al. use PRO-seq to generate highresolution genomic maps of all active RNA polymerases and identify direct transcriptional targets of BET inhibitors within 15 min of treatment. They find that KIT and MCL1 inhibitors are potential therapeutic targets for combination therapies for acute myeloid leukemia along with BET inhibitors.

Highlights

- BET inhibitors impair the release of promoter-proximal paused RNA polymerases
- BET inhibitors repress KIT transcription through its downstream enhancer
- BET inhibitors repress MCL1-targeting microRNAs MIR29C and MIR29B2
- Inhibition of MCL1 triggers cell death in BET-inhibitor-treated AML cells

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High-Resolution Mapping of RNA Polymerases Identifies Mechanisms of Sensitivity and Resistance to BET Inhibitors in t(8;21) AML

Yue Zhao,¹ Qi Liu,²,³ Pankaj Acharya,¹ Kristy R. Stengel,¹ Quanhu Sheng,² Xiaofan Zhou,⁴ Hojoong Kwak,⁵ Melissa A. Fischer,⁶ James E. Bradner,² Stephen A. Strickland,⁶,⁶ Sanjay R. Mohan,⁶,⁶ Michael R. Savona,⁶,⁶ Bryan J. Venters,⁶ Ming-Ming Zhou,¹⁰ John T. Lis,⁵ and Scott W. Hiebert¹,⁶,ఄゃ

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SUMMARY

Bromodomain and extra-terminal domain (BET) family inhibitors offer an approach to treating hematological malignancies. We used precision nuclear run-on transcription sequencing (PRO-seq) to create high-resolution maps of active RNA polymerases across the genome in t(8;21) acute myeloid leukemia (AML), as these polymerases are exceptionally sensitive to BET inhibitors. PRO-seg identified over 1,400 genes showing impaired release of promoter-proximal paused RNA polymerases, including the stem cell factor receptor tyrosine kinase KIT that is mutated in t(8;21) AML. PRO-seq also identified an enhancer 3' to KIT. Chromosome conformation capture confirmed contacts between this enhancer and the KIT promoter, while CRISPRi-mediated repression of this enhancer impaired cell growth. PROseq also identified microRNAs, including MIR29C and MIR29B2, that target the anti-apoptotic factor MCL1 and were repressed by BET inhibitors. MCL1 protein was upregulated, and inhibition of BET proteins sensitized t(8:21)-containing cells to MCL1 inhibition, suggesting a potential mechanism of resistance to BET-inhibitor-induced cell death.

INTRODUCTION

The bromodomain and extra-terminal domain (BET) proteins consist of four family members including BRD2, BRD3, BRD4, and BRDT (Wu and Chiang, 2007). These BET proteins bind to

the acetylated lysines of histone tails and other non-histone nuclear proteins through two conserved N-terminal bromodomains (Dey et al., 2003; Filippakopoulos et al., 2012; Wu and Chiang, 2007; Zhang et al., 2013). BET proteins are typically associated with enhancers and bind to positive transcription elongation factor b (P-TEFb, which contains cyclin T and CDK9), which is critical for the release of promoter-proximal paused RNA polymerases into productive elongation (Chapuy et al., 2013; Liao et al., 1995; Lovén et al., 2013; Marshall and Price, 1995; Peng et al., 1998; Yang et al., 2005). The binding of BRD4 appears to activate P-TEFb by releasing it from the inhibitory HEXIM1-7SK complex (Bartholomeeusen et al., 2012; Jang et al., 2005; Liu et al., 2014). This stimulates CDK9-dependent phosphorylation of the C-terminal domain of RNA polymerase II (RNAPII) as well as negative elongation complexes, DSIF and NELF, which causes dissociation of NELF and switches DSIF into a positive elongation factor, to trigger RNAPII elongation (Wada et al., 1998a, 1998b; Yamaguchi et al., 1999). Small molecule inhibitors of BET proteins, such as JQ1, I-BET, and MS417, mimic the acetylated lysine moiety and competitively bind to the two bromodomains (BD1 and BD2) to displace BET proteins from chromatin (Filippakopoulos et al., 2010; Nicodeme et al., 2010; Zhang et al., 2012a). BET inhibitors show efficacy in preclinical models of acute myeloid leukemia (AML), multiple myeloma, and certain types of lymphoma as well as other cancer types (Chapuy et al., 2013; Dawson et al., 2011; Delmore et al., 2011; Feng et al., 2014; Filippakopoulos et al., 2010; Lockwood et al., 2012; Ott et al., 2012; Zuber et al., 2011).

Consistent with BRD4 interacting with P-TEFb, gene expression studies showed that BET inhibitors induced downregulation of mRNAs including key oncogenes important for cell cycle progression, such as *MYC* and *E2F1*, genes that control cell death such as *BCL2*, as well as lineage-specific oncogenes such as *BCL6* (Chapuy et al., 2013; Dawson et al., 2011; Delmore et al.,



¹Department of Biochemistry, Vanderbilt University School of Medicine, Nashville, TN 37232, USA

²Center for Quantitative Sciences, Vanderbilt University School of Medicine, Nashville, TN 37232, USA

³Department of Biomedical Informatics, Vanderbilt University School of Medicine, Nashville, TN 37232, USA

⁴Department of Biological Sciences, Vanderbilt University, Nashville, TN 37212, USA

⁵Department of Molecular Biology and Genetics, Cornell University, Ithaca, NY 14853, USA

⁶Department of Medicine, Vanderbilt University School of Medicine, Nashville, TN 37232, USA

⁷Department of Medical Oncology, Dana-Farber Cancer Institute, Boston, MA 02115, USA

⁸Vanderbilt-Ingram Cancer Center, Vanderbilt University School of Medicine, Nashville, TN 37232, USA

⁹Department of Molecular Physiology and Biophysics, Vanderbilt University School of Medicine, Nashville, TN 37232, USA

¹⁰Department of Structural and Chemical Biology, Icahn School of Medicine at Mount Sinai, New York, NY 10029, USA

^{*}Correspondence: scott.hiebert@vanderbilt.edu



2011; Ott et al., 2012; Zuber et al., 2011). Genomic binding studies revealed that these genes are associated with BRD4-enriched enhancers that are essential for the efficient transcription of these genes. The so-called "super-enhancers," clusters of enhancers, are particularly sensitive to BET inhibitors causing the selective transcriptional repression of those super-enhancer-driven genes (Chapuy et al., 2013; Lovén et al., 2013). However, as a global chromatin reader, BRD4 is also highly enriched at active promoters, and the mechanism of action of BET inhibitors has been inferred from chromatin immunoprecipitation sequencing (ChIP-seq) studies showing less RNAPII associated with the body of the gene after treatment of BET inhibitors, but these ChIP studies have relatively low resolution and sensitivity, and they do not provide directional information (i.e., the direction polymerase is transcribing) (Lovén et al., 2013; Zhang et al., 2012b).

In these early studies of BET inhibitors, the t(8;21) cell line Kasumi-1 showed an exceptional response to these compounds with 2- to 5-fold higher sensitivity than other leukemia types (Zuber et al., 2011). The t(8,21) is one of the most common chromosomal translocations in AML and yields an immature myeloid leukemia (M2 subtype) that not only expresses the stem cell factor receptor, KIT, but up to 48% of these leukemia also contain activating mutations of KIT (Gao et al., 2015; Park et al., 2011; Paschka et al., 2006; Wang et al., 2005). In fact, KIT mutation is associated with poor outcome in t(8:21) AML and is often observed in relapsed t(8;21) patients (Park et al., 2011; Paschka et al., 2006; Schnittger et al., 2006). Moreover, these cells are sensitive to KIT tyrosine kinase inhibitors, indicating that mutant KIT is an oncogenic mutation in t(8;21) AML (Wang et al., 2005). In total, KIT is activated by point mutation and/or amplified in up to 8% of AML (Forbes et al., 2015; Wheeler et al., 2013). Moreover, KIT is amplified in multiple tumor types including 16% of prostate cancers, 10% of glioblastoma, and 4%-5% of lung cancer (The Cancer Genome Atlas cBioPortal for Cancer Genomics) (Barbieri et al., 2012; Brennan et al., 2013; Cancer Genome Atlas Research Network, 2008, 2012; Cerami et al., 2012; Gao et al., 2013).

We used high-resolution mapping of active RNA polymerases to define the mechanism of transcriptional control by BET inhibitors in t(8;21) AML and find that in 64% of the affected genes, BET inhibitors increased promoter-proximal RNA polymerase pausing. We also defined the effects of the drugs on enhancertemplated RNA (eRNA) and microRNA production. One of the genes suppressed was KIT, which could be a key target of BET inhibitors in 80% of AML. Conversely, BET inhibitors transcriptionally induced the expression of the anti-apoptotic BCL2-family member MCL1, while suppressing microRNAs that regulate the production of MCL1, such that the combined effect was rapid and sustained induction of MCL1. MCL1 expression is associated with drug resistance in multiple myeloma and is upregulated during leukemic relapse (Derenne et al., 2002; Kaufmann et al., 1998; Wuillème-Toumi et al., 2005). While treatment of t(8;21) cells with a selective MCL1 inhibitor had little effect, pretreatment with a BET inhibitor caused a concomitant loss of BCL2 and induction of MCL1, which sensitized these cells to MCL1 inhibition and apoptosis. Thus, BET inhibitors may be extremely useful in AML containing KIT driver mutations, and combination therapy with an MCL1 inhibitor may be beneficial. Moreover, changes in KIT cell surface expression may be useful in monitoring the response to BET inhibitors in clinical trials.

RESULTS

BET Inhibitors Cause Promoter-Proximal Pausing of RNA Polymerases

Early studies identified AML as especially sensitive to inhibitors of BET family members, and the t(8;21) cell line Kasumi-1 appeared to be the most sensitive cell line (Zuber et al., 2011). We extended these results to the t(8;21)-containing SKNO-1 cell line that requires granulocyte-macrophage colony stimulating factor (GM-CSF) for growth (Matozaki et al., 1995) and found that Kasumi-1 and SKNO-1 cells were more sensitive than MOLM13 and MV4-11 when using alamarBlue assays to assess cell metabolism as a surrogate for cell proliferation (Figure S1A; Y.Z., T.M. Heaster, K.R.S., M.C. Skala, M.R.S., and S.W.H., unpublished data). We also extended this work to a more potent BET inhibitor, MS417 (Zhang et al., 2012a), which was 2- to 3-fold more efficacious than JQ1 at restricting Kasumi-1 cell growth (Figure S1A). Surprisingly, when we tested whether BET inhibitors triggered apoptosis in Kasumi-1 cells, both JQ1 and MS417 only had a minor effect in the first 24-48 hr, whereas SKNO-1 showed more cell death at 48 hr (Y.Z., T.M. Heaster, K.R.S., M.C. Skala, M.R.S., and S.W.H., unpublished data; see Figure 7F). Thus, while t(8;21)-containing cells were very sensitive, BET inhibitors inhibited cell proliferation without causing widespread apoptosis.

Kasumi-1 cells are an excellent model of t(8;21) leukemia in terms of epigenetic and transcriptional control, as direct comparison with primary patient samples yielded similar epigenetic marks and transcription factor (TF) occupancy (Ptasinska et al., 2012, 2014). Because Kasumi-1 cells are also extremely sensitive to BET inhibitors, we used precision nuclear run-on transcription coupled with deep sequencing (PRO-seq) to probe the mechanistic basis for this exceptional response to BET inhibitors and to gain insights into the mechanism of action of these compounds (Kwak et al., 2013). PRO-seq provides both directional information and near nucleotide resolution of the genome-wide positions of actively engaged RNA polymerases and is the ideal method to test the proposed mechanism of action of BRD2, BRD3, and BRD4, which are thought to stimulate paused RNA polymerase to elongate via association with P-TEFb (Jang et al., 2005; Yang et al., 2005).

We reasoned that the regulation of nascent transcription should be detectable within the first 1–3 hr after drug treatment and that these early times should reflect the direct transcriptional effects of BET inhibitors, rather than secondary or compensatory effects that might occur 4–8 hr after addition of the drug. Therefore, we initially treated cells with JQ1 or MS417 for 1 and 3 hr, and PRO-seq was performed. At the same time, RNA was collected for RNA-seq analysis to compare the cytoplasmic pools of mRNA with the effects on transcription. The effects of BET inhibitors on transcriptionally engaged RNA polymerases were calculated using the read densities in the promoter-proximal region versus the gene body, and a "pausing index" was defined as the ratio of promoter-proximal density divided by gene body density (Core et al., 2008; Kwak et al., 2013; Min

et al., 2011). Polymerases near the transcription start site (TSS), but moving away from the gene body (i.e., divergent transcripts), were excluded. Importantly, such pausing indices can be compared between samples without normalization, because it measures the relative polymerase content within the same gene (Figure 1A). Since JQ1 and MS417 are different inhibitors targeting the same BET proteins, these two samples were treated as biological replicates, and only genes identified by both inhibitors were used for further analyses.

By comparing pausing indices between control and BETinhibitor-treated samples, we identified 1,905 RefSeq genes showing an increased pausing index 1 hr after treatment, while only 234 genes showed a decreased pausing index (Figures 1B and 1C). The high percentage of overlap between JQ1- and MS417-treated samples verified the accuracy of these studies. When the read counts of the 1,905 genes showing an increase in pausing index 1 hr after BET inhibitor treatment were plotted relative to the TSS (Figure 1D), most of these genes showed a gain of RNA polymerase near the TSS (e.g., MYC, Figures S1C and 1D, upper bracket). However, there was a small group of genes that showed increased levels of promoter-proximal and gene body transcription (yet the net effect was a gain of pausing index; Figure 1D, bottom bracket), suggesting that BET proteins may also play a repressive role in transcription for these genes (e.g., MCL1, Figure S1C). As expected, many genes identified 1 hr after treatment showed an even greater level of pausing at 3 hr after treatment, and more genes were identified with increased pausing indices 3 hr after treatment, which can be classified as delayed-early genes (Figures 1B and 1C). This could be due to a delayed action of BET inhibitors, but MYC protein levels were reduced by about 50% by 3 hr after treatment (Figure S1B), so we were likely assessing indirect or compensatory changes at this time point.

A smaller number of genes (~9%) displayed reduced levels of both promoter-proximal and gene body read densities (Figure 1D, middle bracket), but with a larger loss of gene body density such that these genes were captured in our informatics analysis (e.g., BCL2, Figures 1G and S1C), which could be due to changes in transcription initiation. To examine this aspect of BET regulation in more detail, we directly compared normalized RNA polymerase gene body densities and found 1,033 genes repressed and 346 genes activated by BET inhibitors (Figure 1E, left panels). Among the 1,033 repressed genes, 454 genes showed significantly increased pausing index, which was captured by the pausing index comparison analysis (Figure 1E, right) and is indicative of an inhibition of the release of RNAPII to productive elongation. Interestingly, 575 genes showed no change of pausing index and 4 genes showed decreased pausing index, suggesting that BET inhibitors also affect transcription initiation.

To better trace the kinetics of RNA polymerase elongation after BET inhibitor treatment, we performed PRO-seq with JQ1 treatment for 15 and 30 min and confirmed that our identified target genes were quickly affected by JQ1 at the pause release and/or initiation stages (Figures S1D-S1F). When longer genes were plotted, the gap downstream of the promoter-proximal site revealed the block of paused RNAPII release to productive elongation, which lengthened with time (Figures 1F and 1G). This indicates that BET inhibitors alter pausing release without

affecting the RNA polymerases that have already passed the pausing checkpoint.

We also performed RNA-seq experiments in parallel to better understand the sensitivity of PRO-seq at detecting immediate transcriptional changes (Figures S1G and S1H). 1 hr after treatment, only 187 genes were significantly downregulated at least 1.5-fold in the RNA-seq dataset, while 100 genes were upregulated (Figure S1G). Thus, PRO-seq detected over 5-fold more genes showing repressed gene body transcription (1,033 versus 187). Gene ontology analyses indicated that seven TFs that regulate cell cycle progression and cell proliferation were affected in both PRO-seq and RNA-seq (Figures S1I and S1J), suggesting that by 3 hr we likely detected indirect effects of the de-regulation of these TFs. Indeed, by 3 hr 1,049 mRNAs were downregulated at least 1.5-fold, but only 494 of these showed a direct transcriptional effect of BET inhibition by PRO-seq (Figure S1I). Of note, BRD3 was expressed at low levels so BET inhibitors act only through BRD2 and BRD4 in Kasumi-1 cells. Taken together, PRO-seq enabled us to identify the transcriptional effects of BET inhibitors within the first hour of treatment, ruling out most secondary effects from longer treatment and post-transcriptional regulation.

KIT Is Targeted by BET Inhibitors

In our gene ontology analysis, stress response genes were among the most robustly affected gene clusters (Figure S2A), which is consistent with this group of genes being regulated by paused RNAPII release to productive elongation (Lis et al., 2000; Mahat et al., 2016). In addition, BET inhibitors downregulated genes, including those controlling metabolism (e.g., oxidative phosphorylation, RNA transport, Ribosome biogenesis, and mitochondrial diseases; Figure S2B). This group of genes may have been missed by RNA-seq at early time points as most of these genes have abundant mature mRNA. TF enrichment analyses identified binding motifs in these BET inhibitor-regulated genes for MYC, E2F, IRF, NFMUE1, ELK1, and HIF1 (Figure S2C).

We also noted a large number of genes with increased pausing indices at both time points that are important for hematopoietic malignancies such as DNMT3A, BCL6, IKZF1, ATRX, ETV6, LMO2, CSF3R, PAX5, and TET2 (Figure S2A). In the context of the t(8;21), a gene that was of particular interest was the stem cell factor receptor tyrosine kinase, KIT, which was quickly repressed by BET inhibitors as detected by PRO-seq (Figures 1G, bottom, and 2A). KIT is expressed in 80% of AML and is activated by an N822K mutation in both Kasumi-1 and SKNO-1 cells (Becker et al., 2008; Ikeda et al., 1991; Larizza et al., 2005). Also, this mutant KIT allele is amplified in Kasumi-1 cells and confers sensitivity to the tyrosine kinase inhibitor imatinib (Larizza et al., 2005; Wang et al., 2005). Consistent with the PRO-seq analysis, RNA-seq detected a \sim 50% decrease in KIT mRNA 3 hr after BET inhibitor treatment (Figure 2B). KIT cell surface expression was reduced by JQ1 in Kasumi-1 and SKNO-1 cells beginning at 9 hr after JQ1 treatment (not shown) with more pronounced loss by 24 hr (Figures 2C and 2D). By contrast, the t(8;21) fusion protein RUNX1-ETO was only modestly affected by BET inhibitors (Figure S2D). As expected, treatment of Kasumi-1 cells with the tyrosine kinase inhibitor imatinib also reduced cell growth with an EC₅₀ at \sim 400 nM (Figure 2E, arrow). Combined treatment



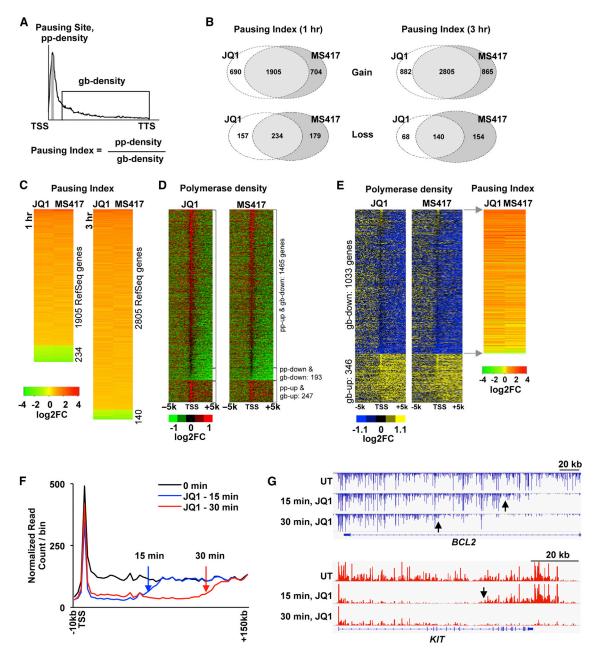


Figure 1. PRO-Seq Analysis of Kasumi-1 Cells Treated with BET Inhibitors for 1 and 3 hr

(A) Illustration of the calculation of RNA polymerase promoter-proximal pausing. TSS, transcription start site; TTS, transcription termination site; pp, promoter-proximal; gb, gene body.

- (B) Venn diagrams show the number of genes displaying an increase (top) or decrease (bottom) of polymerase pausing 1 hr (left) or 3 hr (right) after treatment with 250 nM JQ1 or 125 nM MS417.
- (C) Heatmaps displaying the genes identified by both JQ1 and MS417 treatment showing an increase or decrease of polymerase pausing. Genes were ranked based on log2-transformed fold change of pausing indices.
- (D) Heatmaps displaying log2-transformed fold change (log2FC) of read counts in 200-bp bins \pm 5 kb around the TSSs of the 1,905 genes that show an increase of pausing index after BET inhibitor treatment for 1 hr.
- (E) Heatmaps displaying log2-transformed fold change of read counts in 200-bp bins ± 5 kb around the TSSs of the genes that were up- or downregulated in the gene body by BET inhibitor treatment for 1 hr. Genes were ranked based on log2-transformed fold change of RNA polymerase in the promoter-proximal region. (F) PRO-seq was performed using 250 nM JQ1 at 15- and 30-min time points. Normalized average counts of RNA polymerase were plotted –10 kb to +150 kb from TSSs with 3-kb bins for the genes downregulated by BET inhibitors in the gene body and longer than 150 kb.
- (G) PRO-seq IGV BedGraph screenshots of BCL2 (top) and KIT (bottom) 15 and 30 min after JQ1 treatment. The arrows indicate the interface when RNAPII encountered a block of transcription elongation.

See also Figure S1.

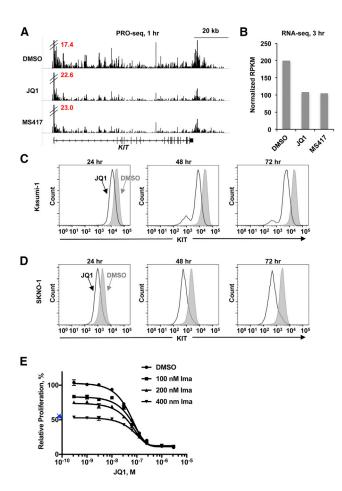


Figure 2. KIT Is Targeted by BET Inhibitors in t(8;21) AML Cell Lines

(A) PRO-seq IGV BedGraph screenshots of *KIT* at 1 hr after BET inhibitor treatment. Double hatch marks indicate that the actual peak size of the paused promoter-proximal RNA polymerases is higher and the numbers are read densities (read per base pair) of polymerase peaks.

(B) Mature mRNA levels were monitored using RNA-seq in the same samples when performing PRO-seq experiments. Read quantification displays an about 50% reduction of *KIT* mRNA expression 3 hr after BET inhibitor treatment.

(C and D) Kasumi-1 (C) and SKNO-1 (D) cells were treated with 250 nM JQ1 for 24, 48, and 72 hr and KIT cell surface expression was quantitatively assessed by flow cytometry. Representative plots are shown (n = 3). Shaded area represents DMSO, and white plots represent JQ1.

(E) Kasumi-1 cells were treated with combined increasing doses of JQ1 and imatinib (Ima) for 3 days, and cell growth was measured by alamarBlue assays. See also Figure S2.

with increasing doses of JQ1 and imatinib enhanced the inhibition of cell growth, suggesting that repression of *KIT* contributes to the effect of BET inhibitors, but that loss of KIT is not the sole mediator of this effect (Figure 2E). Thus, BET inhibitors turn off not only *MYC* but also drivers of leukemogenesis such as mutant *KIT*, which likely contributes to the sensitivity of t(8;21)-containing cells to BET inhibitors.

BET Inhibitors Affect eRNA Transcription

BET family members are typically associated with enhancers and are displaced from chromatin by BET inhibitors, which

is commonly associated with enhancer inactivation (Chapuy et al., 2013; Lovén et al., 2013). Because active enhancers produce 5'-capped eRNAs and the activation of eRNAs correlates with increased transcription of neighboring genes (Andersson et al., 2014; Core et al., 2014), PRO-seq data allow identification of transcriptionally active enhancers, whereas traditional ChIP-seq experiments of histone modifications (e.g., H3K27ac) cannot distinguish active from inactive enhancers.

We first assessed the myeloid-specific *MYC* super-enhancer (Shi et al., 2013). Consistent with the ChIP-seq data, divergent transcription identified specific transcriptional start sites associated with the five previously identified "sub-enhancers" (Figure 3A; Figure S3A). In addition, we identified a sixth enhancer in this region containing a well-defined transcriptional start site (E6 in Figures 3A and S3A). We also identified a previously unrecognized potential long eRNA being expressed from E5 (Figures 3A, S3A, and S3B). Importantly, normalized read densities were substantially reduced by BET inhibitors, especially within the second, fifth, and sixth enhancers (Figure 3B), which is consistent with *MYC* being regulated by this super-enhancer in myeloid malignancies.

Next, we performed ChIP-exonuclease sequencing (ChIPexo) for H3K27ac and analyzed the PRO-seq data in comparison with ChIP-exo data for H3K27ac in Kasumi-1 cells to identify enhancers. Only intergenic enhancers were identified in this study, as transcripts derived from intragenic enhancers are difficult to distinguish from gene transcription. With a method modified from a previously published algorithm (Hah et al., 2013), we identified 2,631 active intergenic enhancers, which were also marked with H3K27ac (Figure 3C). When compared with H3K27ac marked intergenic regions, PRO-seq was able to identify weak enhancers with higher sensitivity, and it distinguished active enhancers from open chromatin marked by H3K27ac (Figure S3C). Using a 1.5-fold change as a cutoff, 613 enhancers had fewer RNA polymerases associated after treatment with either JQ1 or MS417, while only 93 enhancers were activated (Figure 3D). With a simplified proximity rule (Lovén et al., 2013), which assigned enhancers to their closest active genes within a 50-kb window, we identified 208 active genes that were associated with repressed enhancers (Figure 3E, left), and their pausing indices were significantly increased by BET inhibitors (Figure 3F).

Previous enhancer-gene association studies relied on mature mRNA levels, while PRO-seq can directly measure transcriptional changes using gene body read counts. We selected transcriptionally active intergenic enhancers that are marked by overlapping H3K27ac ChIP-exo peaks, BRD4 ChIP-seq peaks previously reported in myeloid cells, and PRO-seq defined enhancers (Dawson et al., 2014; Gröschel et al., 2014; Poss et al., 2016). Although both BRD2 and BRD4 are highly expressed in Kasumi-1 cells (data not shown), we focused on BRD4 as it best correlates with JQ1 genomic target sites (Anders et al., 2014). Next, we identified so-called super-enhancers that have multiple H3K27ac peaks (Figure S3D) and examined their function on regulating neighboring gene transcriptional changes. We consistently found that genes associated with super-enhancers showed greater reduction of transcription as measured by PRO-seq gene body read count (Figure S3E).



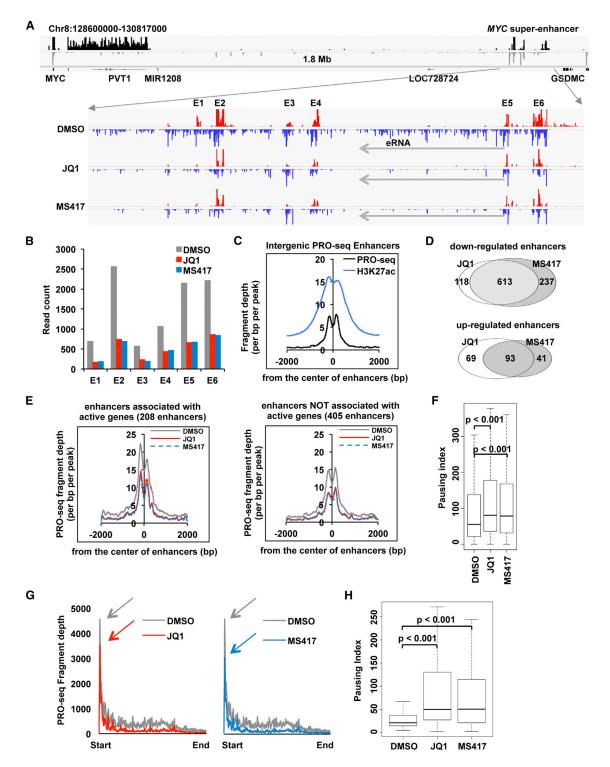


Figure 3. PRO-Seq Identification of Enhancers and Long eRNA Affected by BET Inhibitors

(A and B) MYC super-enhancer, which consists of six sub-enhancers (E1–E6), is located 1.8 Mb downstream of the MYC gene (A, top). Gray arrows indicate eRNA transcription initiated from E5, which was repressed by BET inhibitors (A, bottom). Quantification of eRNA synthesis from the six sub-enhancers, showing that BET inhibitors affect eRNA transcription 1 hr after drug addition (B).

(legend continued on next page)

⁽C) PRO-seq called intergenic enhancers showing an enrichment of H3K27ac. PRO-seq double peaks around the center of enhancers indicate divergent transcription.

⁽D) Venn diagrams showing the number of enhancers repressed or activated at 1 hr after BET inhibitor treatment.

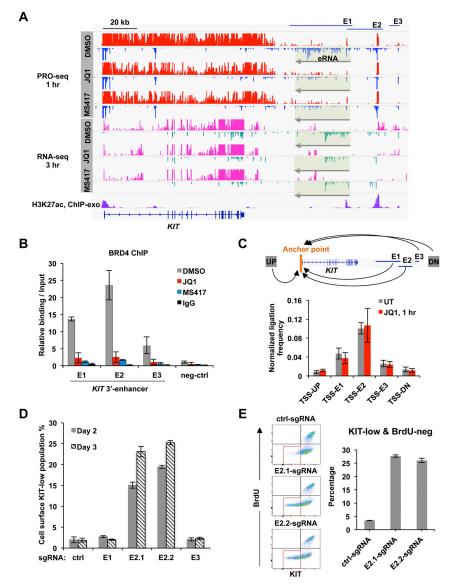


Figure 4. PRO-Seq Identifies a KIT Enhancer

(A) IGV BedGraph screenshots of the enhancer 3' to KIT. Top three PRO-seq gene tracks display three sub-enhancers (E1-E3) identified by bidirectional transcription. Green shadows with arrows highlight the eRNA transcribed from E1 and repressed by BET inhibitors, which is confirmed by RNA-seq (middle tracks). The bottom ChIP-exo gene track shows the enrichment of H3K27ac at the transcription initiation sites of the three sub-enhancers

(B) ChIP assays show BRD4 enrichment around the transcription initiation sites of the three sub-enhancers E1-E3, which was dramatically reduced 1 hr after BET inhibitor treatment. Neg-ctrl represents a negative control region that shows no transcriptional activity determined by PRO-seq. Data are mean \pm SEM (n = 3).

(C) 3C experiments were performed with an anchor point fixed at the KIT promoter. Transcriptionally inactive regions up- (UP) and downstream (DN) of the KIT gene were used as negative controls. All ligation efficiencies were normalized to digested and religated BAC templates. Data are mean \pm SE (n = 3).

(D and E) The KIT downstream sub-enhancers were repressed by CRISPRi with individual sgRNAs targeting E1, E2 (two different sgRNAs E2.1 and E2.2 are shown), and E3. KIT cell surface expression was analyzed by flow cytometry 2 and 3 days after infection. Bar graphs show the quantification of cells expressing low levels of KIT. Data are mean \pm SE (n = 3) (D). The cells with low KIT expression were pulsed with BrdU on day 3 for cell cycle analysis. Representative plots on the left display no BrdU incorporation (BrdU-neg) in the KIT-low cell population (red boxes), which is quantified and shown in bar graphs on the right. Data are mean \pm SE (n = 3) (E).

We also found 70 PRO-seg defined downregulated enhancers spanning larger genomic regions (>10 kb), and 47 of these enhancers contain at least 2 sub-enhancers characterized by bi-directional transcription. A major advantage of PRO-seq is that it allows the identification of eRNAs, and we noted that 62 of the 70 large enhancers transcribed at least 1 long eRNA (6 kb~46 kb) from their corresponding initiation sites (e.g., an intergenic enhancer region transcribing long eRNAs is shown in Figure S3F). BET inhibitors reduced the release of RNA polymer-

ases into the eRNA body, causing an increase in the pausing index for these eRNAs (Figures 3G and 3H).

See also Figure S4.

PRO-Seq Identifies a KIT Enhancer 3' to KIT

Importantly, this analysis identified a region 50 kb 3' to the KIT locus containing three actively transcribed regions and a long eRNA, which was dramatically affected by BET inhibitors (Figure 4A, gray arrows in the top three tracks). This eRNA was also captured by our RNA-seq analysis using polyA enriched

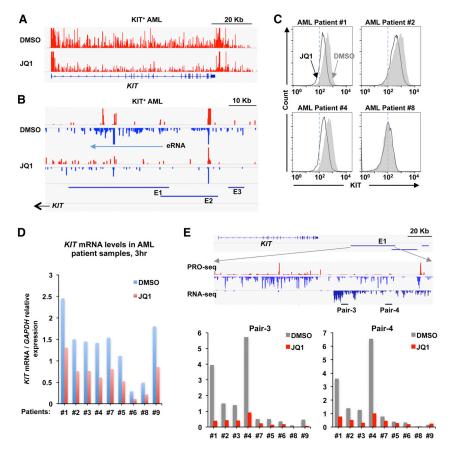
⁽E) Repressed enhancers associated with (left) or without (right) a proximal active gene.

⁽F) Active genes associated with repressed enhancers display an increase of pausing index at 1 hr after treatment with BET inhibitors. p values were calculated using one-sided Wilcoxon signed-rank test.

⁽G) Average RNA polymerase levels within the 62 long eRNAs that are affected by BET inhibitor treatment, showing the loss of eRNA transcription elongation. Arrows indicate the peaks of RNA polymerases.

⁽H) Box plots showing pausing indices of the 62 long eRNAs upon BET inhibitor treatment. p values were calculated using one-sided Wilcoxon Signed-Rank test. See also Figure S3.





RNA and RT-PCR using primers throughout the length of this eRNA (Figure 4A, gray arrows in the middle three tracks; Figure S4A). Interestingly, the eRNA detected by RNA-seq was shorter (12 kb) than the transcribed region, and its 3' end is followed by an accumulation of RNAPII in the PRO-seg analysis. which is commonly observed 3' of the transcription cleavage and polyadenylation site associated with RNA processing (Figure 4A) (Core et al., 2008; Glover-Cutter et al., 2008). The three sub-enhancers (identified by divergent transcription) were enriched with H3K27ac in ChIP-exo analysis, which is commonly used as histone markers for active enhancers (Figure 4A, bottom track). Moreover, ChIP assays revealed BRD4 binding around the transcription initiation sites of the three sub-enhancers (E1, E2, and E3), which was almost completely disrupted 1 hr after BET inhibitor treatment (Figure 4B).

To validate that this region is a KIT enhancer, we performed chromosome conformation capture (3C) experiments and showed that E2 interacts with the KIT promoter (Figure 4C). This interaction was not affected by JQ1 treatment, suggesting that BET proteins are not required for chromatin looping. We next used a CRISPRi system to repress the individual sub-enhancers, in which a deactivated Cas9 was fused to a transcriptional repression domain KRAB and can be directed to a genomic region targeted by sgRNA. Multiple sgRNAs were designed for each sub-enhancer, and only repressing E2 reduced KIT expression (Figures 4D and S4B). Moreover, the cells with reduced KIT cell surface expression showed impaired cell cycle progression, as

Figure 5. PRO-seq Analysis of Primary AML **Patient Cells**

(A and B) Primary AML patient cells were treated with DMSO or 250 nM JQ1 for 1 hr. PRO-seq IGV BedGraph screenshots show a repressive effect of JQ1 on KIT transcription (A) and its downstream enhancer and eRNA (B).

(C-E) Analysis of the eRNA transcribed from the KIT enhancer in nine primary AML patient samples treated with DMSO or 250 nM JQ1. Total RNA was collected 3 hr after JQ1 treatment to detect transcriptional changes of KIT mRNA (D) and the 3' eRNA (E). Transcription of eRNA was determined by two pairs of primers targeting the short form of eRNA that can be detected by both PRO-seq and RNA-seq (E, top; also refer to Figure S4A). KIT cell surface expression was measured by flow cytometry 48 hr after JQ1 treatment (C), and representative KIT-positive and -negative AML patient cells are displayed.

See also Figure S5.

no DNA synthesis was detected by BrdU incorporation assays (Figure 4E). Thus, E2 is the predominant enhancer region regulating KIT transcription, which is important

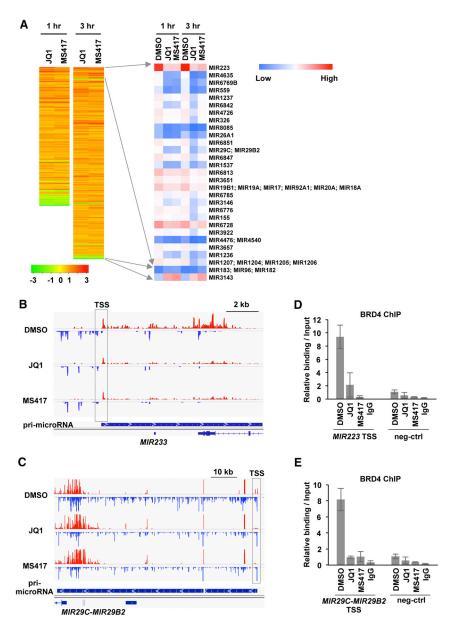
for cell cycle progression.

PRO-Seq Analysis of Primary AML Patient Cells

Consistent with KIT being expressed in hematopoietic stem and progenitor cells and

AML cells, the transcription of KIT and its putative enhancer was only detected in Kasumi-1 cells and was not detectable in other available PRO-seq datasets from lymphoid tissues (Figure S5A), re-enforcing the usefulness of BET inhibitors in KIT-positive AMLs. Therefore, we extended our findings to a primary KITpositive AML patient sample for which ample material was available (PRO-seq requires 20 million cells). PRO-seq analysis confirmed that JQ1 impaired transcription elongation in patient cells, with 2,680 genes showing increased pausing indices, while only 736 genes showed decreased pausing indices (Figure S5B). Among the 3,416 affected genes, 58% of them were also significantly changed by BET inhibitors in Kasumi-1 cells. Consistent with our findings in Kasumi-1 cells, genes with increased pausing indices were important for hematopoietic diseases and cell survival and proliferation (Figures 5A and S5C). We were also able to detect the six enhancers that regulate MYC (Figure S5D) and the three enhancers that regulate KIT (Figure 5B). Moreover, transcription elongation of the KIT eRNA was inhibited by JQ1 (Figure 5B, blue arrow).

Next, we tested whether BET inhibitor treatment affects KIT expression in primary AML patient samples. Nine primary AML samples were treated ex vivo with JQ1 (samples 1, 2, 3, 4, 5, and 7 were KIT positive, while samples 6, 8, and 9 were KIT negative) (Figure 6C). Among the KIT-positive AML samples, five (samples 1, 2, 3, 4, and 7) expressed significant amounts of KIT mRNA and its 3' eRNA, which was dramatically downregulated by JQ1 (Figures 6D and 6E) with concurrent loss of cell



surface KIT expression (Figure 6C). By contrast, the KIT-negative samples (e.g., samples 6 and 8) barely expressed any KIT mRNA or its 3' eRNA (Figures 6D and 6E). Although samples 5 and 10 also expressed detectable KIT mRNA and their mRNA levels were reduced by JQ1, we could not detect any change of KIT cell surface expression for sample 5, and sample 10 failed to express KIT on its cell membrane, suggesting that non-transcriptional mechanisms also contribute to regulating KIT expression in some cases.

Suppression of MicroRNAs Enhance MCL1 Induction by BET Inhibitors

MicroRNAs are small non-coding RNAs (~22 nucleotides) that regulate gene expression by affecting mRNA stability and translation (Ha and Kim, 2014). About 60% of microRNAs are co-tran-

Figure 6. BET Inhibitors Suppress MicroRNAs **Targeting MCL1**

(A) Heatmaps on the left display the pri-microRNAs identified by both JQ1 and MS417 treatment showing an increase or decrease of RNA polymerase pausing index. Each cell represents log2-transformed fold change of pausing indices and they are ranked based on fold change of pri-microRNA body densities. A zoomed view of the top 20% most changed pri-microRNAs both 1 and 3 hr after BET inhibitor addition (27 with increased pausing indices and 2 with decreased pausing indices) is shown on the right, which displays pri-microRNA body densities. MicroRNAs transcribed from individual primicroRNAs are listed on the side.

(B and C) PRO-seq IGV BedGraph screenshots of primicroRNAs transcribing MIR223 (B) and MIR29C; MIR29B2 (C). Gray boxes indicate pri-microRNA TSSs defined by divergent transcription.

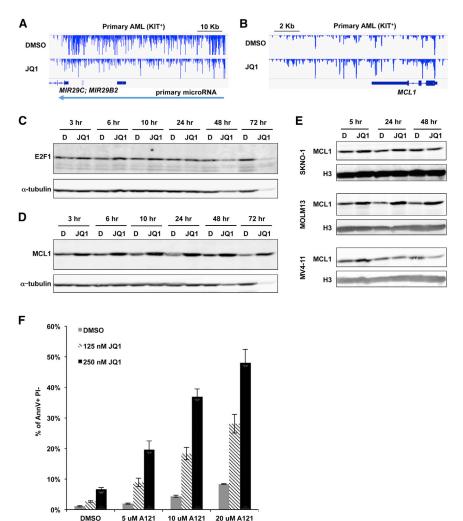
(D and E) Kasumi-1 cells were treated with DMSO, 250 nM JQ1, and 125 nM MS417 for 1 hr, ChIP assays show the enrichment of BRD4 binding around the TTSs of MIR223 (D) and MIR29C-MIR29B2 (E) pri-microRNAs, which was dramatically reduced by BET inhibitors. Neg-ctrl represents a negative control region, which shows no transcriptional activity determined by PRO-seq data. Data are mean \pm SEM (n = 3)

See also Figure S6.

scribed from protein-coding gene,s and another 40% of them are localized to intergenic regions. MicroRNAs are initially transcribed as long primary microRNAs (pri-microRNAs) and undergo sequential processing steps during maturation (Ha and Kim, 2014). While the functional roles of microRNAs have been extensively characterized, it has been challenging to define their transcriptional regulation without accurately predicting the TSSs of their pri-microRNAs. By examining all transcriptionally engaged RNA polymerases, PROseq allowed us to identify 787 pri-micro-RNAs (206 intergenic and 581 intragenic

pri-microRNAs) expressed in Kasumi-1 cells. By 1 hr after BET inhibitor addition, 126 pri-microRNAs showed an increase of pausing index, while only 15 pri-microRNAs showed a decrease of pausing index (Figure 6A, left). 3 hr after BET inhibitor addition, 189 pri-microRNAs were detected with an increased pausing index and 7 pri-microRNAs showed a decreased pausing index (Figure 6A, left). Among the 126 pri-microRNAs showing an increase of pausing index after 1 hr of treatment, the highly expressed pri-microRNA for MIR223, which targets E2F1, was the most affected by BET inhibitors (Figures 6A, right, 6B, and S6). Moreover, the pri-microRNA of MIR29B2 and MIR29C that target MCL1 was among the top 20% of the pri-microRNAs with gain of pausing (Figures 6A, right, 6C, and S6). ChIP assays further revealed BRD4 binding at the TSSs of the pri-microRNAs of MIR223 and MIR29B2-MIR29C, which was dramatically





reduced 1 hr after BET inhibitor treatment, indicating that *MIR223*, *MIR29B2*, and *MIR29C* are direct targets of BET inhibitors (Figures 6D and 6E).

Inhibition of BET Proteins Sensitizes Cells to MCL1 Inhibition

BET-inhibitor-mediated loss of expression of MIR29B2 and MIR29C was of special interest, because our PRO-seq analysis identified MCL1 as transcriptionally activated by BET inhibitors (Figures S1C, S7A, and S7B). Expression of MCL1 would be expected to counterbalance the loss of BCL2 expression to suppress apoptosis. PRO-seq analysis of the AML patient sample confirmed the reduction of MIR29B2 and MIR29C in AML blasts and the activation of MCL1 transcription (Figures 7A and 7B). When we analyzed protein levels of E2F1 and MCL1, E2F1 was only slightly increased at early times by JQ1, whereas MCL1 was upregulated within 3 hr treatment of BET inhibitors (Figures 7C and 7D). Thus, at early time points after BET inhibitor addition, the effects are more likely due to increased transcription. Consistent with Kasumi-1 cells being arrested in the G_0/G_1 phase by BET inhibitors (Figure S1A; Y.Z., T.M. Heaster,

Figure 7. BET Inhibitors Sensitize Kasumi-1 Cells to MCL1 Inhibition

(A and B) PRO-seq IGV BedGraph screenshots display reduced transcription of *MIR29B2* and *MIR29C* and activated *MCL1* transcription after JQ1 treatment in primary AML patient cells.

(C and D) Time-course examination of E2F1 (C) and MCL1 (D) protein levels in Kasumi-1 cells treated with DMSO (D) and 250 nM JQ1. α -tubulin was used as loading control. The downregulation of α -tubulin at 48 and 72 hr reflects an effect of cell cycle arrest by long-term treatment of BET inhibitors.

(E) SKNO-1 cells were treated 250 nM JQ1, and MOLM13 and MV4-11 cells were treated with 500 nM JQ1. DMSO (D) served as control. MCL1 protein levels were assessed by western blot.

(F) Kasumi-1 cells were pre-treated with DMSO, 125 nM, and 250 nM JQ1 for 2 days, and the MCL1 selective inhibitor A1210477 (A121) was added to cell culture for another 6 hr before Annexin V (AnnV) and propidium iodide (PI) staining. Apoptotic cell population was detected by AnnV positivity and the absence of PI staining. Data are mean \pm SEM (n = 4).

K.R.S., M.C. Skala, M.R.S., and S.W.H., unpublished data), E2F1 protein levels subsided at 48–72 hr after BET inhibitor addition (Figure 7C). In contrast, the upregulation of MCL1 was sustained at higher levels throughout the entire 3-day treatment (Figure 7D). This was true not only in Kasumi-1 cells, but in other cell types such as MOLM-13 that also showed less cell death, whereas MV4-11 failed to upregulate MCL1 and the cells were

killed by JQ1 (Figure 7E). The induction of MCL1 provides a possible mechanism of resistance to cell death even in the face of declining *BCL2* levels. Therefore, we tested the combined effects of inhibiting BET proteins and MCL1. Whereas Kasumi-1 cells were not overtly sensitive to A1210477, a selective MCL1 inhibitor (Leverson et al., 2015), pretreating these cells with JQ1 to reduce the levels of *BCL2* sensitized cells to the effects of the MCL1-selective compound and triggered apoptosis (Figure 7F). Therefore, transcriptional changes of *MIR29C*, *MIR29B2*, *MCL1*, and *KIT* can be used to monitor patient outcomes after BET inhibitor treatment and provide molecular basis for combination therapy with MCL1 inhibitors.

See also Figure S7.

DISCUSSION

While there is justifiable excitement about the therapeutic efficacy of BET inhibitors in AML, only a small portion of the cell types tested was affected by these compounds, and in most cases it is unclear what causes sensitivity or resistance (Chapuy et al., 2013; Delmore et al., 2011; Filippakopoulos et al., 2010; Lockwood et al., 2012; Ott et al., 2012; Zuber et al., 2011). PRO-seq creates a

high-resolution and high-sensitivity map of all active RNA polymerases and provides directionality of transcription not available using ChIP-seq (Kwak et al., 2013). This analysis not only showed a rapid loss of RNA polymerase elongation at MYC but also identified an eRNA produced by a sub-enhancer within the MYC super-enhancer that was affected by BET inhibitors. Moreover, PRO-seq analysis identified KIT as a transcriptional target of BET inhibitors. This receptor tyrosine kinase is expressed in 80% of AML and is activated by mutations in \sim 8% of AML, and activated KIT is a cooperating mutation in up to 48% of t(8;21) AML (Forbes et al., 2015; Ikeda et al., 1991; Paschka et al., 2006; Wang et al., 2005). This suppression of KIT expression by BET inhibitors provided increased sensitivity to inhibitors of KIT (Figure 2E). Thus, regulation of KIT provides rationale for the clinical use of BET inhibitors in KIT mutated or amplified AML, and it could be of use in many other types of AML that express KIT. In this regard, it is noteworthy that KIT is activated or amplified in prostate, breast, colon, gastric cancer, as well as other solid cancers, which might indicate the utility of BET inhibitors in many types of cancer (Cerami et al., 2012; Gao et al., 2013).

In addition to KIT, we noted many other genes that are involved in hematopoietic malignancies displayed a gain of promoter-proximal pausing or impaired elongation in Kasumi-1 cells treated with BET inhibitors. Most notable were a group of genes that regulate transcriptional elongation including MLLT1 (ENL), MLLT3 (AF9), MLLT6 (AF17), MLLT10 (AF10), ELL, MLL, Cyclin T2 (CCNT2, a component of P-TEFb), BRD2, WDR5, and DOT1L. This might suggest that there is a feedback loop in which BET proteins regulate components of the super elongation complex. We also noted that multiple members of the mediator complex (MED12, MED20, MED24, MED25, and MED30) were affected by BET inhibitors, as were the chromatin remodeling factors INO80, INO80D, CBP, GCN5 (KAT2A), and KAT6B. This implies that BET inhibitors have even more dramatic effects the longer that the cells are exposed to the compound. It also highlights specific potential roles for BET inhibitors in AML.

BRD4 is preferentially enriched at enhancers and super-enhancers, and much attention has focused on the selective effect of BET inhibitors on super-enhancer-regulated genes (Chapuy et al., 2013; Lovén et al., 2013). We were able to use the divergent transcription that is associated with transcription initiation to identify defined transcriptional start sites within active enhancers and super-enhancers that were affected by BET inhibitors, as well as identify eRNAs. While the mechanism of action of BET proteins is typically envisioned to be due to enhancers looping to promoters to stimulate transcriptional elongation through BET proteins contacting P-TEFb, when coupled with recent work on the architecture of enhancers (Core et al., 2014), our study suggests that generation of functional eRNAs is an attractive model for how enhancers regulate tissue-specific gene expression. If eRNAs expressed at an enhancer regulate elongation at promoters, one might expect that the effect of BET inhibitors on eRNA expression would precede effects at promoters. However, PRO-seq at just 15 min after BET inhibitor addition was not able to temporally separate the effect of BET inhibitors at the MYC super-enhancer from MYC mRNA production (data not shown), leaving open the question of how enhancers function to control tissue-specific gene expression and the contribution of BET family members to the communication between enhancers and promoters.

A major advantage of PRO-seq is that it allows the assessment of rapid changes in transcription from all RNA polymerases and all regions of the genome at one time, including microRNAs. While there is rapid loss of MYC expression in cells treated with BET inhibitors (Chapuy et al., 2013; Delmore et al., 2011; Ott et al., 2012; Zuber et al., 2011), the loss of MIR223 transcription, which regulates E2F1, could force some tumor cells to continue to cycle. Indeed, we noted that E2F1 levels only declined when the cells entered into a G₀/G₁ arrest at later time points (Figure 7). Likewise, the loss of transcription of the pri-microRNA for MIR29B2 and MIR29C, which targets the anti-apoptotic gene MCL1, likely contributed to the accumulation of MCL1 at later time points. While most genes were repressed by BET inhibitors, these compounds stimulated the transcription of MCL1 (Figure S1). Given that BRD4 associates with the MCL1 promoter (data not shown) (Lovén et al., 2013), one can speculate that BRD4 acts to maintain the transcription of MCL1 under normal conditions, but the absence of BRD4 upon BET inhibitor treatment allows MCL1 transactivation by other TFs that act through different mechanisms. When coupled with the loss of MIR29B2 and MIR29C, MCL1 induction likely allows cells to resist apoptosis. The development of selective MCL1 inhibitors that can be used in combination with BET inhibitors, which suppress KIT, MYC, and BCL2 expression, may prove to be a potent combination for attacking KIT-positive AML in the clinic.

EXPERIMENTAL PROCEDURES

PRO-Seq Library Preparation and Data Analysis

Nuclear run-on assays were performed and sequencing libraries were constructed as described in the Supplemental Experimental Procedures (Kwak et al., 2013). Libraries were submitted to the Vanderbilt Technologies for Advanced Genomics (VANTAGE) for sequencing. Pre-processed reads were aligned to the human genome hg19 (downloaded from UCSC) using Bowtie2 (v.2.2.4) (Langmead and Salzberg, 2012). See Supplemental Experimental Procedures for additional methods and statistical and informatics analysis.

RNA-Seq

PolyA+ RNA was enriched for library preparation and submitted to VANTAGE for RNA sequencing. Pre-processed reads were aligned to the human transcriptome hg19 (downloaded from UCSC) using TopHat (v.2.0.10) (Kim et al., 2013). See Supplemental Experimental Procedures for statistical and informatics analysis.

Flow Cytometry for KIT Expression

Cells were seeded at 0.2×10^6 /ml and treated with DMSO or 250 nM JQ1 for 24, 48, and 72 hr. 0.5×10^6 cells were collected at each time point, washed with cold PBS, and stained with antibodies against KIT (catalog no. 313204, BioLegend) at 4° C for 15 min before flow cytometry. All flow cytometry figures were generated using Flowjo.

Assessment of Apoptosis

Kasumi-1 cells were seeded at 0.2×10^6 /ml and immediately treated with DMSO, 125 nM JQ1, and 250 nM JQ1 for 2 days. Then increasing doses of A1210477 were added to the pre-treated cells for an additional 6 hr. Apoptosis was analyzed using a FITC-AnnexinV/PI Apoptosis Detection kit (catalog no. 556547, BD PharMingen).



ChIP Assays

Kasumi-1 cells were seeded at $0.5 \times 10^6/\text{ml}$ at the day of experiment, treated with DMSO, 250 nM JQ1, and 125 nM MS417 for 1 hr. DNA and protein was crosslinked with 1% formaldehyde at room temperature for 10 min, and crosslinking was terminated by 125 mM glycine. Crosslinked cells were washed twice with cold PBS, lysed with 1% SDS lysis buffer, and sonicated. 100 μ l of sonicated chromatin was transferred to 900 μ l cold dilution buffer, BRD4 antibody (Bethyl Laboratories) and protein G magnetic beads were added, and reactions were incubated at 4°C overnight. Immunoprecipitated protein and DNA complexes were washed, and crosslinking was reversed before DNA isolation. 4 μ l of DNA was used in each quantitative PCR to assess BRD4 enrichment. Data were calculated relative to inputs and a transcriptionally inactive region for normalization and background reduction.

Statistical Analyses

The significance of pausing index change for each gene upon treatment was evaluated using Fisher's exact test followed by multiple testing adjustment. Wilcoxon signed-rank test was used for paired comparisons of RNA polymerase pausing indices between two treatments. p values or FDR < 0.05 were considered as statistically significant. For quantifications of apoptosis and BRD4 binding, results were presented as mean \pm SEM. Statistical analyses were also provided in each corresponding figure legend. See Supplemental Experimental Procedures for additional statistical and informatics analysis.

Supplemental Experimental Procedures

A more robust and detailed description of the methods used and associated references are listed in the Supplemental Experimental Procedures.

ACCESSION NUMBERS

The accession number for the PRO-seq, RNA-seq, and ChIP-exo data is GEO: GSE83660.

SUPPLEMENTAL INFORMATION

Supplemental Information includes Supplemental Experimental Procedures and seven figures and can be found with this article online at http://dx.doi.org/10.1016/j.celrep.2016.07.032.

AUTHOR CONTRIBUTIONS

Y.Z. performed the experiments and bioinformatics analyses, prepared the figures, and wrote the manuscript; Q.L., Q.S., and X.Z. assisted with the bioinformatics studies; P.A., K.S., and M.A.F. assisted the performance of experiments; H.K. and J.T.L. assisted PRO-seq experiments and contributed to the writing; J.E.B. provided JQ1; M.-M.Z. provided MS417; S.A.S., S.R.M., and M.R.S. provided primary AML patient cells; B.J.V. assisted with the ChIP-exo experiments; and S.W.H. supervised the study and wrote the manuscript.

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