

Supplemental Information for ‘Pioneer factor GAF cooperates with PBAP and NURF to regulate transcription’

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This file contains all code used to analyze data presented in ‘Pioneer factor GAF cooperates with PBAP and NURF to regulate transcription’.

The goal of this document is reproducibility: This is the exact code used to generate the figures in the final version of the paper, and any user should be able to replicate our analyses starting from raw data (GSE149339 or PRJNA627972).

Supplementary Code 1. PRO-seq alignment pipeline

This pipeline can be found here:

http://github.com/jaj256/PROseq_alignment.sh

Analysis in this paper was performed using commit 55a08db

```
#!/bin/bash

#####
# Tue Mar  5 14:04:36 EST 2019                                     #
# This is a pipeline script for handling paired end               #
# PRO-seq data with UMIs on both ends of the read.               #
# Run this script in a directory that has one folder             #
# named "fastq" which contains the data.                         #
# Fastq files must have identical names other than               #
# ending in _R1.fastq and _R2.fastq.                             #
#####

## Parameters
THREADS=50 # Threads to use for multithreaded applications
UMI_LEN=6  # Length of UMI in basepairs

## UMI Flags (set to Y or N as appropriate)
FIVEP_UMI="Y" # Is there a UMI on the 5' end of the read?
THREEP_UMI="Y" # Is there a UMI on the 3' end of the read?

## Adaptor sequences to clip. Default = Tru-Seq small RNA
ADAPTOR_1="TGGAATTCTCGGGTGCCAAGGAAGTCCAGTCAC"
ADAPTOR_2="GATCGTCGGACTGTAGAACTCTGAACGTGTAGATCTCGGTGGTCGCCGTATCATT"

## Genomes. Fill in paths.
GENOME_EXP="/home/jaj256/genome/dm6/dm6Hsp70AaOnly"
GENOME_SPIKE="/home/jaj256/genome/dm6hg38/dm6hg38" ## USE REPEAT MASKED VERSION!!
SPIKE_PREFIX="hg38" ## This is the prefix you've used on your spike in chromosomes
RDNA="/home/jaj256/genome/dm3hg38/dm3hg38rDNA"

## Mapq value for filtering multimappers
MAPQ=10

#####
#                               PIPELINE                           #
#####

# Unzipping if needed
echo "unzipping..."
for FILE in fastq/*
do
    if [[ "$FILE" == *.gz ]]
    then
        gunzip $FILE &
    fi
done
```

```

wait

# Removing extra info from filenames.
# This is general and works with files from Cornell BRC.
# If filenames are formatted differently, does nothing.
echo "renaming if needed..."
for FILE in $(ls fastq/)
do
    NEW=fastq/"$(echo "$FILE" |
        sed 's/^[0-9]\+_ [0-9]\+_ [0-9]\+_ [0-9A-Z]\+_// ' |
        sed 's/_[ATCG]\{6,8\}_/_/ ')"
    if [ ! -s "$NEW" ]
    then
        mv fastq/"$FILE" "$NEW"
    fi
done

mkdir -p logs
mkdir -p logs/fastqc

# Running fastqc on files
echo "running fastqc if needed..."
for FILE in fastq/*.fastq
do
    if [ ! -s logs/fastqc/"$(basename ${FILE/.fastq/_fastqc.zip})" ]
    then
        fastqc "$FILE" -o logs/fastqc --quiet &
    fi
done
wait

mkdir -p trimmedFastq

# Autodetecting paired end files
echo "detecting paired end files..."
NUM=$(ls fastq | wc -l)
NUM_REDUCED=$(ls fastq | sed 's/_R.*//' | uniq | wc -l)
if [[ $NUM == $NUM_REDUCED ]]
then
    PAIRED="N"
    echo "detected ""$NUM"" single end fastq files. exiting..."
    exit
else
    PAIRED="Y"
    echo "detected ""$NUM_REDUCED"" paired end fastq files"
fi

# Trimming adapters and filtering rRNA reads
# Four logical branches for 5' and 3' UMI, 5' only, 3' only, and no UMI
echo "trimming adapters and filtering rDNA reads..."
mkdir -p logs/fastp
mkdir -p logs/rRNA

```

```

mkdir -p trimmedFastq
if [[ $PAIRED == "Y" ]]
then
  # Branches for either 3' UMI or both UMIs
  if [[ $THREEP_UMI == "Y" ]]
  then
    # Branch for both UMIs
    if [[ $FIVEP_UMI == "Y" ]]
    then
      for PAIR in $(ls fastq | sed 's/_R[1-2].*//' | uniq )
      do
        if [ ! -s trimmedFastq/${PAIR}_R1.fastq ]
        then
          echo "trimming adapters and filtering rRNA reads for "${PAIR}"
          (fastp \
          -i fastq/${PAIR}_R1.fastq \
          -I fastq/${PAIR}_R2.fastq \
          --adapter_sequence $ADAPTOR_1 \
          --adapter_sequence_r2 $ADAPTOR_2 \
          --umi \
          --stdout \
          --umi_loc=per_read \
          --umi_len=${UMI_LEN} \
          --html logs/fastp/${PAIR}_fastp.html \
          -w $(echo ${THREADS}/3 | bc)\
          -c \
          --overlap_len_require 15 2> logs/fastp/${PAIR}_fastp.log) |
          (bowtie2 \
          --fast-local \
          --un-conc trimmedFastq/${PAIR}.fastq \
          --interleaved - \
          -x ${RDNA} \
          --threads $(echo ${THREADS}/3*2 | bc)
          2> logs/rRNA/${PAIR}_rRNA_bowtie.log) > /dev/null
        fi
      done
    # Branch for just 3' UMI
    else
      for PAIR in $(ls fastq | sed 's/_R[1-2].*//' | uniq )
      do
        if [ ! -s trimmedFastq/${PAIR}_R1.fastq ]
        then
          echo "trimming adapters and filtering rRNA reads for "${PAIR}"
          (fastp \
          -i fastq/${PAIR}_R1.fastq \
          -I fastq/${PAIR}_R2.fastq \
          --adapter_sequence $ADAPTOR_1 \
          --adapter_sequence_r2 $ADAPTOR_2 \
          --umi \
          --stdout \
          --umi_loc=read1 \
          --umi_len=${UMI_LEN} \
          --html logs/fastp/${PAIR}_fastp.html \

```

```

-w $(echo ${THREADS}/3 | bc) \
-c \
--overlap_len_require 15 2> logs/fastp/${PAIR}_fastp.log) |
(bowtie2 \
--fast-local \
--un-conc trimmedFastq/${PAIR}.fastq \
--interleaved - \
-x ${RDNA} \
--threads $(echo ${THREADS}/3*2 | bc)
2> logs/rRNA/${PAIR}_rRNA_bowtie.log) > /dev/null
fi
done
fi
# Branch for only 5' UMI or no UMIs
else
# Branch for only 5' UMI
if [[ $FIVEP_UMI == "Y" ]]
then
for PAIR in $(ls fastq | sed 's/_R[1-2].*//' | uniq)
do
if [ ! -s trimmedFastq/${PAIR}_R1.fastq ]
then
echo "trimming adapters and filtering rRNA reads for "${PAIR}"
(fastp \
-i fastq/${PAIR}_R1.fastq \
-I fastq/${PAIR}_R2.fastq \
--adapter_sequence $ADAPTOR_1 \
--adapter_sequence_r2 $ADAPTOR_2 \
--umi \
--stdout \
--umi_loc=read2 \
--umi_len=${UMI_LEN} \
--html logs/fastp/${PAIR}_fastp.html \
-w $(echo ${THREADS}/3 | bc) \
-c \
--overlap_len_require 15 2> logs/fastp/${PAIR}_fastp.log) |
(bowtie2 \
--fast-local \
--un-conc trimmedFastq/${PAIR}.fastq \
--interleaved - \
-x ${RDNA} \
--threads $(echo ${THREADS}/3*2 | bc)
2> logs/rRNA/${PAIR}_rRNA_bowtie.log) > /dev/null
fi
done
# Branch for no UMI
else
for PAIR in $(ls fastq | sed 's/_R[1-2].*//' | uniq)
do
if [ ! -s trimmedFastq/${PAIR}_R1.fastq ]
then
echo "trimming adapters and filtering rRNA reads for "${PAIR}"
(fastp \

```

```

-i fastq/${PAIR}_R1.fastq \
-I fastq/${PAIR}_R2.fastq \
--adapter_sequence $ADAPTOR_1 \
--adapter_sequence_r2 $ADAPTOR_2 \
--stdout \
--html logs/fastp/${PAIR}_fastp.html \
-w $(echo ${THREADS}/3 | bc) \
-c \
--overlap_len_require 15 2> logs/fastp/${PAIR}_fastp.log) |
(bowtie2 \
--fast-local \
--un-conc trimmedFastq/${PAIR}.fastq \
--interleaved - \
-x ${RDNA} \
--threads $(echo ${THREADS}/3*2 | bc)
2> logs/rRNA/${PAIR}_rRNA_bowtie.log) > /dev/null
fi
done
fi
fi
fi

# Cleaning up filenames in trimmedFastq (bowtie automatically names PE --un output)
for FILE in trimmedFastq/*1.fastq
do
    if [ ! -s ${FILE}/.1.fastq/_R1.fastq ]
    then
        mv "$FILE" ${FILE}/.1.fastq/_R1.fastq
    fi
done

for FILE in trimmedFastq/*2.fastq
do
    if [ ! -s ${FILE}/.2.fastq/_R2.fastq ]
    then
        mv "$FILE" ${FILE}/.2.fastq/_R2.fastq
    fi
done

# Aligning to spike in genome to get normalization factors
mkdir -p spikeBAM
mkdir -p logs/spikeAlign

if [[ "$PAIRED" == "Y" ]]
then
    for PAIR in $(ls trimmedFastq | sed 's/_R[1-2].*//' | uniq)
    do
        if [ ! -s "spikeBAM/${PAIR}_hg38.BAM" ]
        then
            echo "aligning ${PAIR} to spike in genome"
            (bowtie2 \
            --local \
            --very-sensitive-local \

```

```

--threads $(echo ${THREADS}/3*2 | bc) \
--no-unal \
--no-mixed \
--no-discordant \
-x "$GENOME_SPIKE" \
-1 "trimmedFastq/${PAIR}_R1.fastq" \
-2 "trimmedFastq/${PAIR}_R2.fastq" \
2> logs/spikeAlign/${PAIR}_spikeAlign.log) |
samtools view -hS -f 2 -q ${MAPQ} |
perl -n -e 'print $_ if (/^@/ || /'${SPIKE_PREFIX}'/ ) ' |
samtools view -b |
samtools sort -@ $(echo ${THREADS}/3 | bc) -o spikeBAM/${PAIR}.BAM
samtools index spikeBAM/${PAIR}.BAM
fi
done
fi

# Aligning to experimental genome
mkdir -p BAM
mkdir -p logs/align

if [[ "$PAIRED" == "Y" ]]
then
for PAIR in $(ls trimmedFastq | sed 's/_R[1-2].*//' | uniq)
do
if [ ! -s "BAM/${PAIR}.BAM" ]
then
echo "aligning ${PAIR} to experimental genome"
(bowtie2 \
--local \
--sensitive-local \
--threads $(echo ${THREADS}/3*2 | bc) \
-x "$GENOME_EXP" \
-1 "trimmedFastq/${PAIR}_R1.fastq" \
-2 "trimmedFastq/${PAIR}_R2.fastq" \
2> logs/align/${PAIR}_align.log) |
samtools view -bS -f 2 -q ${MAPQ} |
samtools sort -@ $(echo ${THREADS}/3 | bc) -o BAM/${PAIR}.BAM
samtools index BAM/${PAIR}.BAM
fi
done
fi

# Deduplicating with UMIs (experimental BAM)
mkdir -p BAMdeDuped
mkdir -p logs/deDup

for FILE in BAM/*.BAM
do
if [ ! -s "BAMdeDuped/$(basename ${FILE%.BAM}_deDuped.BAM)" ]
then
(umi_tools dedup \
-I "$FILE" \

```



```

        --umi-separator=":" \
        --paired \
        -S "BAMdeDuplicated/$(basename ${FILE%.BAM}_deDuplicated.BAM)" \
    )> "logs/deDup/$(basename ${FILE%.BAM}_deDup.log)" &&
    samtools index "BAMdeDuplicated/$(basename ${FILE%.BAM}_deDuplicated.BAM)"
fi
done

# Deduplicating with UMIs (Spike-In BAM)
mkdir -p spikeBAMdeDuplicated
mkdir -p logs/spikededup

for FILE in spikeBAM/*.BAM
do
    if [ ! -s "spikeBAMdeDuplicated/$(basename ${FILE%.BAM}_deDuplicated.BAM)" ]
    then
        (
            umi_tools dedup \
            -I "$FILE" \
            --paired \
            --umi-separator=":" \
            -S "spikeBAMdeDuplicated/$(basename ${FILE%.BAM}_deDuplicated.BAM)" \
        )> "logs/spikededup/$(basename ${FILE%.BAM}_deDup.log)" &&
        samtools index "spikeBAMdeDuplicated/$(basename ${FILE%.BAM}_deDuplicated.BAM)"
    fi
done

# Generating table of Alignment metrics
mkdir -p info

if [ ! -s info/infoTable.tsv ]
then
touch info/infoTable.tsv
echo -e Name'\t'\
        RawReads'\t'\
        NonDimerReads'\t'\
        %dimer'\t'\
        insertSize'\t'\
        rRNAreads'\t'\
        %rRNA'\t'\
        passedFilters'\t'\
        bowtieConcordant'\t'\
        bowtieMulti'\t'\
        bowtieUnal'\t'\
        bowtieOverallMap%\t'\
        bowtieConcordant%\t'\
        bowtieMulti%\t'\
        bowtieUnal%\t'\
        uniqueMapped'\t'\
        uniqueMappedNondup'\t'\
        %PCRDups'\t'\
        uniqueMappedSpikein'\t'\

```

```

uniqueMappedSpikeinNondup'\t'\
spikeInPCRDups% >> info/infoTable.tsv

for SAMPLE in $(ls BAM/*.BAM | sed 's/.BAM//' | sed 's/BAM\\\\/' )
do
NAME=${SAMPLE}
RAW_READS=$(cat logs/fastp/${SAMPLE}_fastp.log |
grep "total reads:" | head -n 1 |
awk '{print $3}')
TRIMMED_READS=$(cat logs/fastp/${SAMPLE}_fastp.log |
grep "total reads:" | tail -n 1 |
awk '{print $3}')
PER_DIMER=$(echo "(1-${TRIMMED_READS}/${RAW_READS})*100" | bc -l)%
INSERT_SIZE=$(cat logs/fastp/${SAMPLE}_fastp.log |
grep "Insert size peak" |
awk '{print $8}')
PASSED_FILTERS=$(cat logs/align/${SAMPLE}_align.log |
grep "reads; of these:" |
awk '{print $1}')
RRNA=$(echo ${TRIMMED_READS}-${PASSED_FILTERS} | bc )
PER_RRNA=$(echo ${RRNA}/${RAW_READS}*100 | bc -l)%
B_CONC=$(cat logs/align/${SAMPLE}_align.log |
grep "aligned concordantly exactly 1 time$" |
awk '{print $1}')
B_MULTI=$(cat logs/align/${SAMPLE}_align.log |
grep "aligned concordantly >1 times$" |
awk '{print $1}')
B_UNAL=$(cat logs/align/${SAMPLE}_align.log |
grep "aligned concordantly 0 times$" |
awk '{print $1}')
B_OAP=$(cat logs/align/${SAMPLE}_align.log |
grep "overall alignment rate$" |
awk '{print $1}')
B_CONC_PER=$(echo ${B_CONC}/${PASSED_FILTERS}*100 | bc -l)%
B_MULTI_PER=$(echo ${B_MULTI}/${PASSED_FILTERS}*100 | bc -l)%
B_UNAL_PER=$(echo ${B_UNAL}/${PASSED_FILTERS}*100 | bc -l)%
UNIQ_MAPPED=$(cat logs/deDup/${SAMPLE}_deDup.log |
grep "Input Reads:" | awk '{print $10}')
UNIQ_MAPPED_DEDUP=$(cat logs/deDup/${SAMPLE}_deDup.log |
grep "Number of reads out:" | awk '{print $8}')
PER_DUPS=$(echo "(1-${UNIQ_MAPPED_DEDUP}/${UNIQ_MAPPED})*100" | bc -l)%
UNIQ_MAPPED_SPIKE=$(cat logs/spikededup/${SAMPLE}_deDup.log |
grep "Input Reads:" | awk '{print $10}')
UNIQ_MAPPED_DEDUP_SPIKE=$(cat logs/spikededup/${SAMPLE}_deDup.log |
grep "Number of reads out:" | awk '{print $8}')
PER_DUPS_SPIKE=$(echo "(1-${UNIQ_MAPPED_DEDUP_SPIKE}/${UNIQ_MAPPED_SPIKE})*100" |
bc -l)%

echo -e $NAME'\t'\
$RAW_READS'\t'\
$TRIMMED_READS'\t'\
$PER_DIMER'\t'\
$INSERT_SIZE'\t'\

```

```

$RRNA'\t\'
$PER_RRNA'\t\'
$PASSED_FILTERS'\t\'
$B_CONC'\t\'
$B_MULTI'\t\'
$B_UNAL'\t\'
$B_OAP'\t\'
$B_CONC_PER'\t\'
$B_MULTI_PER'\t\'
$B_UNAL_PER'\t\'
$UNIQ_MAPPED'\t\'
$UNIQ_MAPPED_DEDUP'\t\'
$PER_DUPS'\t\'
$UNIQ_MAPPED_SPIKE'\t\'
$UNIQ_MAPPED_DEDUP_SPIKE'\t\'
$PER_DUPS_SPIKE >> info/infoTable.tsv

done
fi

# Making non-normalized bigWig files
mkdir -p bw
for FILE in BAMdeDuplicated/*.BAM
do
    if [ ! -s "bw/${basename ${FILE}/.BAM/_fwd.bw}" ]
    then
        bamCoverage \
        --bam $FILE \
        --skipNonCoveredRegions \
        --outFileName bw/${basename ${FILE}/.BAM/_fwd.bw} \
        --binSize 1 \
        --numberOfProcessors ${THREADS} \
        --normalizeUsing None \
        --Offset 1 \
        --samFlagInclude 82
    fi
    if [ ! -s "bw/${basename ${FILE}/.BAM/_rev.bw}" ]
    then
        bamCoverage \
        --bam $FILE \
        --skipNonCoveredRegions \
        --outFileName bw/${basename ${FILE}/.BAM/_rev.bw} \
        --binSize 1 \
        --numberOfProcessors ${THREADS} \
        --normalizeUsing None \
        --Offset 1 \
        --samFlagInclude 98
    fi
done

```

Supplementary Code 2. ATAC-seq alignment pipeline

```
#!/bin/bash
# This script is for aligning/peak calling/making bigwig signal tracks for ATAC-seq data

# Number of threads to use for applications that support multithreading
THREADS=50

# Unzipping if needed
for FILE in fastq/*
do gunzip $FILE -q &
done

# Renaming files
for FILE in $(ls fastq/)
do
    NEW=fastq/"$(echo "$FILE" |
    sed 's/^[0-9]\+_[0-9]\+_[0-9]\+_[0-9A-Z]\+/_/' |
    sed 's/_[ATCG]\{6,8\}_/_/')"
    if [ ! -s "$NEW" ]
    then
        mv fastq/"$FILE" "$NEW"
    fi
done

# Running Fastqc
mkdir -p logs
mkdir -p logs/fastqc

for FILE in fastq/*.fastq
do
    if [ ! -s logs/fastqc/"$(basename ${FILE/.fastq/_fastqc.zip})" ]
    then
        fastqc "$FILE" -o logs/fastqc/ --quiet &
    fi
done
wait

# Aligning files
mkdir -p BAM
mkdir -p logs/align

for PAIR in $(ls fastq | sed 's/_R[1-2].*//' | uniq)
do
    if [ ! -s "BAM/${PAIR}.BAM" ]
    then
        (bowtie2 --local --very-sensitive-local --threads ${THREADS} \
        --no-unal -I 10 -X 1000 -x ~/genome/dm6/dm6Hsp70AaOnly \
        -1 fastq/${PAIR}_R1.fastq \
        -2 fastq/${PAIR}_R2.fastq \
        2> logs/align/${PAIR}_align.log) |
        samtools view -bS -q 10 -f 2 |
        samtools sort -o BAM/${PAIR}.BAM
    fi
done
```

```

        samtools index BAM/${PAIR}.BAM
    fi
done

# Plotting coverage of reads < 120 bp insert (ATAC Hypersensitivity)
mkdir -p bwDHS

for FILE in BAM/*.BAM
do
    bamCoverage \
        --bam ${FILE} \
        --outFileName bwDHS/$(basename ${FILE}/.BAM/_AtacDHS.bw) \
        --binSize 1 \
        --numberOfProcessors ${THREADS} \
        --normalizeUsing None \
        --skipNAs \
        --extendReads \
        --maxFragmentLength 120
done

# Plotting coverage of centers of reads 200 > insert > 130 (mononucleosomes)
mkdir -p bwMonoNucs
for FILE in repMergedBAM/*.BAM
do
    if [ ! -s bwMonoNucs/$(basename ${FILE}/.BAM/_AtacMonoNucs.bw) ]
    then
        bamCoverage \
            --bam ${FILE} \
            --outFileName bwMonoNucs/$(basename ${FILE}/.BAM/_AtacMonoNucs.bw) \
            --binSize 1 \
            --MNase \
            --numberOfProcessors ${THREADS} \
            --normalizeUsing None
    fi
done

# Calling peaks on all BAM files (pooling all samples)
source /programs/bin/util/setup_macs2.sh

mkdir -p peaks
macs2 callpeak \
    -t BAM/*.BAM \
    -f BAMPE \
    -g dm \
    -n "ALL_ATAC_PEAKS" \
    --outdir peaks \
    --call-summits

```

Supplementary Code 3. 3' RNA-seq alignment pipeline

```
#!/bin/bash

# Number of threads to use for applications that support multithreading
THREADS=50

# building STAR index
/programs/STAR/STAR \
  --runThreadN ${THREADS} \
  --runMode genomeGenerate \
  --genomeDir /workdir/jaj256/gaf/new/RNAseq/STARindex \
  --genomeFastaFiles ~/genome/dm6hg38ERCC/dm6hg38ERCC.fa \
  --sjdbGTFfile ~/genome/dm6hg38ERCC/dm6hg38ERCC.gtf \
  --sjdbOverhang 68

mkdir -p trimmedFastq
mkdir -p logs
mkdir -p logs/fastp

# Trimming Adapters and polyA sequences, and extracting UMIs
for FILE in fastq/*.fastq
do
  fastp \
    -i ${FILE} \
    -o trimmedFastq/${(basename ${FILE})} \
    --length_required 25 \
    --adapter_fasta adapters.fa \
    --umi \
    --umi_loc=read1 \
    --umi_len=6 \
    --html logs/fastp/${(basename ${FILE}/.fastq/_fastp.html)} \
    -w ${THREADS} 2> logs/${(basename ${FILE}/.fastq/_fastp.log)}
done

# Aligning to the combined experimental/Spike-in genome
mkdir -p BAM
mkdir -p logs/STAR
for FILE in trimmedFastq/*.fastq
do
  /programs/STAR/STAR \
    --runThreadN ${THREADS} \
    --genomeDir STARindex/ \
    --readFilesIn ${FILE} \
    --outFilterType BySJout \
    --outFilterMultimapNmax 20 \
    --alignSJoverhangMin 8 \
    --alignSJDBoverhangMin 1 \
    --outFilterMismatchNmax 999 \
    --outFilterMismatchNoverLmax 0.1 \
    --alignIntronMin 20 \
    --alignIntronMax 1000000 \
```

```

        --alignMatesGapMax 1000000 \
        --outSAMattributes NH HI NM MD \
        --outSAMtype BAM SortedByCoordinate \
        --outFileNamePrefix BAM/$(basename ${FILE/.fastq/}) \
2> logs/STAR/$(basename ${FILE/.fastq/_STAR.log})
done

# Deduplicating using UMIs
mkdir -p BAMdeDuped
mkdir -p logs/deDup
for FILE in BAM/*.BAM
do
    umi_tools dedup \
    -I ${FILE} \
    -S "BAMdeDuped/$(basename ${FILE})" \
    --umi-separator=":" \
    > logs/deDup/$(basename ${FILE%.BAM}_deDup.log) &
done
wait

for FILE in BAMdeDuped/*.BAM
do samtools index $FILE &
done
wait

# Splitting spike-in and experimental alignments
mkdir -p dm6BAM
mkdir -p ERCCBAM

for FILE in BAMdeDuped/*.BAM
do
    samtools view -hS -q 255 -F 4 ${FILE} |
    perl -n -e 'print $_ if (/^\@/ || /ERCC/)' |
    samtools view -b |
    samtools sort -o ERCCBAM/$(basename ${FILE})

    samtools view -hS -q 255 -F 4 ${FILE} |
    perl -n -e 'print $_ if (/^\@/ || !(/hg38/ || /ERCC/))' |
    samtools view -b |
    samtools sort -o dm6BAM/$(basename ${FILE})
done

# Counting # of alignments to experimental and spike-in genomes
mkdir -p spikeCounts
touch spikeCounts/spikeCounts.tsv
echo -e sample"\t"dm6reads"\t"ERCCreads > spikeCounts/spikeCounts.tsv

for FILE in BAMdeDuped/*.BAM;
do
    FILENAME=$(basename ${FILE%.BAM})
    DM6=$(samtools view -c dm6BAM/${FILENAME}.BAM)
    ERCC=$(samtools view -c ERCCBAM/${FILENAME}.BAM)
    echo -e ${FILENAME}"\t"${DM6}"\t"${ERCC} >> spikeCounts/spikeCounts.tsv
done

```

```

done

# Making bw signal tracks (non-normalized)
for FILE in dm6BAM/*.BAM
do
    bamCoverage \
    --bam $FILE \
    --skipNonCoveredRegions \
    --outFileFormat bigwig \
    --outFileName bdg/${basename ${FILE/.BAM/_fwd.bw}} \
    --binSize 1 \
    --numberOfProcessors ${THREADS} \
    --normalizeUsing None \
    --Offset 1 \
    --filterRNAstrand reverse
    bamCoverage \
    --bam $FILE \
    --skipNonCoveredRegions \
    --outFileFormat bigwig \
    --outFileName bdg/${basename ${FILE/.BAM/_rev.bw}} \
    --binSize 1 \
    --numberOfProcessors ${THREADS} \
    --normalizeUsing None \
    --Offset 1 \
    --filterRNAstrand forward
done

```


Supplementary Code 4. CUT&RUN alignment pipeline

```
#!/bin/bash

# Number of threads to use for applications that support multithreading
THREADS=50

# Running Fastqc
mkdir -p logs
mkdir -p logs/fastqc

for FILE in fastq/*.fastq
do
    if [ ! -s logs/fastqc/${basename ${FILE}/.fastq/_fastqc.zip} ]
    then
        fastqc ${FILE} -q -o logs/fastqc &
    fi
done
wait

# Trimming adapters, aligning, sorting, and indexing
mkdir -p BAM
mkdir -p logs/align
mkdir -p logs/fastp

for PAIR in $(ls fastq | sed 's/_R[1-2].*//' | uniq)
do
    if [ ! -s BAM/${PAIR}.BAM ]
    then
        (fastp \
        -i fastq/${PAIR}_R1.fastq \
        -I fastq/${PAIR}_R2.fastq \
        --adapter_sequence AGATCGGAAGAGCACACGTCTGAACTCCAGTCAC \
        --adapter_sequence_r2 \
        AGATCGGAAGAGCGTCGTGTAGGGAAAGAGTGTAGATCTCGGTGGTCGCCGTATCATT \
        --stdout \
        --html logs/fastq/${PAIR}_fastp.log \
        -w $(echo ${THREADS}/3 | bc) \
        -c --overlap_len_require 15
        2> logs/fastp/${PAIR}_fastp.log) |
        (bowtie2 \
        --very-sensitive-local \
        --no-unal --no-discordant \
        --interleaved - \
        -x ~/genome/dm6/dm6Hsp70AaOnly \
        --threads $(echo ${THREADS}/3*2 | bc)
        2> logs/align/${PAIR}_align.log) |
        samtools view -bS -f 2 -q 10 |
        samtools sort -o BAM/${PAIR}.BAM
        samtools index BAM/${PAIR}.BAM
    fi
done
```

```
# Making bigwig files
mkdir -p bw
for FILE in BAM/*.BAM;
do
    if [ ! -s bw/$(basename ${FILE/BAM/bw}) ]
    then
        bamCoverage \
        -b $FILE \
        -o bw/$(basename ${FILE/BAM/bw}) \
        --binSize 10 \
        -p ${THREADS} \
        --normalizeUsing None \
        --skipNonCoveredRegions \
        --extendReads \
        --maxFragmentLength 120
    fi
done
```

Supplementary Code 5. ChIP-seq realignment pipeline

```
#!/bin/bash

# Number of threads to use for applications that support multithreading
THREADS=50

# Aligning
mkdir -p logs/align
mkdir -p BAM
for FILE in fastq/*.fastq
do
    (bowtie2 \
     --very-sensitive-local \
     --no-unal \
     -x ~/genome/dm6/dm6Hsp70AaOnly \
     --threads ${THREADS} \
     -U ${FILE} 2> logs/align/${(basename ${FILE%.fastq})}_align.log) |
    samtools view -bS -q 10 |
    samtools sort -o BAM/${(basename ${FILE%.fastq})}.BAM
    samtools index BAM/${(basename ${FILE%.fastq})}.BAM
done

# Making bigWig files
mkdir -p bw
for FILE in BAM/*.BAM
do
    bamCoverage \
    -b ${FILE} \
    -o bw/${(basename ${FILE%.BAM}.bw)} \
    --binSize 10 \
    --extendReads 200 \
    --normalizeUsing None \
    --skipNonCoveredRegions
done
```

Supplementary Code 6. Data Analysis (R Scripts)

All code chunks in this section were run in RStudio.
See Supplementary Code 7 for version of all packages used.

Loading packages

```
library(tidyverse)
library(DESeq2)
library(ggpubr)
library(viridis)
library(scales)
library(rtracklayer)
library(GenomicRanges)
library(BiocParallel)
library(BRGenomics)
library(extrafont)
library(patchwork)
library(plyr)
library(ComplexHeatmap)
library(circlize)
library(tiff)
library(eulerr)
library(gridExtra)
loadfonts()
source("/Users/julius/Google Drive/R/customPackages/browserPlotR.R")
```

Functions

Custom R functions used throughout the remainder of the code chunks

```
fc_corr.jj <-
  function(res1,
           res2,
           pval = 0.01,
           cols = c("#BB0021", "#3B4992", "#BBBBBB")) {
    # This function takes the path to two saved DESeq2 Results files (.tsv)
    # Parses them, separates them into activated and repressed classes (padj < 0.05)
    # and plots a scatter of res1 l2FC by res2 l2FC
    # with a glm fit (shaded CI = 95%) of each class
    # significance calling is based on the pvalue of res1

    res.df <- data.frame(
      res1_l2FC = as.data.frame(res1)$log2FoldChange,
      res1_padj = as.data.frame(res1)$padj,
      res2_l2FC = as.data.frame(res2)$log2FoldChange,
      res2_padj = as.data.frame(res2)$padj
    )

    res.df <- res.df[!is.na(res.df$res1_padj) & !is.na(res.df$res2_l2FC),]
```

```

res.df$sig <- ifelse((res.df$res1_padj < pval),
                    ifelse(res.df$res1_l2FC < 0,
                            "down",
                            "up"),
                    "ns"
)

res.df$sig <- factor(res.df$sig, levels = c("up", "down", "ns"))

num_up <- nrow(filter(res.df, sig == "up"))
num_down <- nrow(filter(res.df, sig == "down"))
max_y <- max(res.df$res2_l2FC)
min_y <- min(res.df$res2_l2FC)
max_x <- max(res.df$res1_l2FC)
min_x <- min(res.df$res1_l2FC)

p <-
  ggplot(res.df,
    aes(x = res1_l2FC, y = res2_l2FC)) +
  geom_abline(intercept = 0, slope = 1) +
  geom_point(
    data = filter(res.df, sig == "ns"),
    size = 0.75,
    alpha = 0.75,
    stroke = 0,
    show.legend = F,
    color = cols[3]
  ) +
  geom_point(
    data = filter(res.df, sig == "up"),
    size = 0.75,
    alpha = 0.5,
    stroke = 0,
    show.legend = F,
    color = cols[1]
  ) +
  geom_point(
    data = filter(res.df, sig == "down"),
    size = 1,
    alpha = 0.75,
    stroke = 0,
    show.legend = F,
    color = cols[2]
  ) +
  geom_smooth(
    data = filter(res.df, sig == "down"),
    method = "glm",
    level = 0.95,
    size = 0.5,
    color = "black",
    show.legend = F
  ) +

```

```

    geom_smooth(
      data = filter(res.df, sig == "up"),
      method = "glm",
      level = 0.95,
      size = 0.5,
      color = "black",
      show.legend = F
    ) +
    xlab(paste0(deparse(substitute(res1)), "log2 FC")) +
    ylab(paste0(deparse(substitute(res2)), "log2 FC")) +
    ggtheme.jj() +
    geom_vline(xintercept = 0, linetype = "dashed", alpha = 0.5)+
    geom_hline(yintercept = 0, linetype = "dashed", alpha = 0.5)+
    annotate(geom = 'text',
      label = num_up,
      x = max_x * 0.9,
      y = min_y * 0.9,
      color = cols[1]
    )+
    annotate(geom = 'text',
      label = num_down,
      x = min_x * 0.9,
      y = max_y * 0.9,
      color = cols[2]
    )

  return(p)
}

subset_DESeq.jj <- function(df,
                           col_data,
                           scale_facts,
                           pos_inc,
                           contrast,
                           formula) {
  # Function that takes a dataframe and calls DESeq2
  # On a subset of the dataframe. Provide a df of counts,
  # a colData object (see DESeq2 Vignette), a vector of
  # scale factors, positions to include as a numeric
  # vector (pos.inc), and a character vector to contrast
  # Returns a DESeq2 results object

  df <- df[, pos_inc]
  col_data <- col_data[pos_inc, , drop = FALSE]
  dds <- DESeqDataSetFromMatrix(countData = df,
                                colData = col_data,
                                design = formula)
  sizeFactors(dds) <- scale_facts[pos_inc]
  dds <- DESeq(dds)
  res <- results(dds,
                 contrast = contrast)

```

```

}

ggMetaplot <- function(meta.mat) {
  # Takes a subsampled matrix (output by metaSubsample in BRGenomics)
  # and produces a 'default' metaplot using ggplot. This can easily be customized
  # using standard ggplot syntax:
  # ggMetaplot(meta.mat) + scale_color_manual(values = c("blue", "red")), for example
  return(
    ggplot(
      meta.mat,
      aes(
        x = x,
        y = mean,
        ymax = upper,
        ymin = lower,
        color = sample.name
      )
    ) +
    geom_line(size = 1, alpha = 1) +
    geom_ribbon(
      alpha = 0.2,
      aes(fill = sample.name),
      color = NA,
      show.legend = T
    ) +
    ggtheme.jj() +
    xlab("Distance from TSS (bp)") +
    ylab("Mean + 75% CI")
  )
}

RPMnorm <- function(gr){
  gr$score <- (1e6 / sum(gr$score)) * gr$score
  return(gr)
}

ggtheme.jj <- function() {
  # Custom theme options for ggplot2 graphics
  theme_classic(base_size=10, base_family="Helvetica") %+replace%
  theme(
    axis.text = element_text(size = 8),
    axis.ticks = element_line(colour = "black"),
    legend.key = element_blank(),
    panel.background = element_rect(fill = "white", colour = NA),
    panel.border = element_blank(),
    panel.grid.major = element_blank(),
    panel.grid.minor = element_blank(),
    strip.background = element_blank(),
    strip.text = element_text(size=10),
    plot.title = element_text(hjust = 0.5, size = 10),
    axis.title = element_text(size = 8, face = "bold")
  )
}

```

```

}

rep_corr_scatter.jj <-
  function(df,
           sep_colnames_into = c(),
           colsep = "_",
           drop_cols = NA
  ) {
    # wrapper function that takes a dataframe of counts, splits them
    # by replicate, and plots density of points in hexbins, rep1 vs rep2
    # df: dataframe of counts. Must contain at least one column of unique IDs.
    #     each column should have counts for one sample
    # sep_colnames_into: character vector of columns to separate colnames into.
    #     one value MUST be "condition", the other must be "replicate".
    #     List NAs for unused spots, according to colsep. For example,
    #     a column named "BAP170_NHS_Rep1" would be
    #     c("condition", NA, "rep")
    # colsep: field separator for above split
    # drop_cols: vector of column numbers or names to drop, useful if starting df
    #     has other info like start, stop, etc
    # Other attributes that ggplots can have can be added by calling this function
    # then using + ylab() + xlab() etc

    if(!is.na(drop_cols)) {
      df <- df[,-drop_cols]
    }

    df.ltb <- df %>%
      gather("colname", "count", -tx_name) %>%
      separate(colname, into = sep_colnames_into, sep = colsep) %>%
      spread(rep, count)

    p <- ggplot(df.ltb, aes(
      x = Rep1,
      y = Rep2
    )) +
      geom_abline(
        intercept = 0,
        slope = 1,
        linetype = "solid",
        size = 0.5,
        alpha = 0.75
      ) +
      geom_hex(bins = c(25, 25)) +
      stat_cor(
        method = "spearman",
        label.x.npc = c("left"),
        label.y.npc = c("top"),
        output.type = "text",
        hjust = -0.1
      ) +
      scale_x_log10() +
      scale_y_log10() +

```



```

    scale_fill_viridis_c(name = "Density") +
    ggtheme.jj()

  return(p)
}

maplot.jj <- function(res, padj_cutoff = 0.01, l2fc_cutoff = 0) {
  # This function takes a DESeq2 Results object and plots a pretty MA plot.
  # res = the results object
  # padj_cutoff = significance cutoff (dbl)
  # l2fc_cutoff = log2 fold change cutoff (dbl)

  resdf <- as.data.frame(res)
  resdf$name <- row.names(resdf)
  resdf[is.na(resdf$padj),]$padj <- 1
  resdf <-
    drop_na(mutate(resdf, "class" = if_else(
      padj < padj_cutoff,
      if_else(log2FoldChange > 0, "Activated", "Repressed"),
      "Unchanged"
    )))

  resdf$class[which(abs(resdf$log2FoldChange) < l2fc_cutoff)] <- "Unchanged"

  numAct <- nrow(filter(resdf, class == "Activated"))
  numRep <- nrow(filter(resdf, class == "Repressed"))
  numUch <- nrow(filter(resdf, class == "Unchanged"))
  max_y <- max(resdf$log2FoldChange)
  min_y <- min(resdf$log2FoldChange)
  max_x <- max(resdf$baseMean)
  min_x <- min(resdf$baseMean)
  x_lim_left <- 10^floor(log10(min_x))
  x_lim_right <- 10^ceiling(log10(max_x))
  y_lim_top <- round_any(max_y, 2, ceiling)
  y_lim_bottom <- round_any(min_y, 2, floor)

  lseq <- function(from, to, length.out) {
    # logarithmic spaced sequence
    # blatantly stolen from library("emdbook"), because need only this
    exp(seq(log(from), log(to), length.out = length.out))
  }

  breaks_x <- lseq(
    from = x_lim_left,
    to = x_lim_right,
    length.out = (log10(x_lim_right) - log10(x_lim_left) + 1)
  )

  breaks_y <- seq(y_lim_bottom, y_lim_top, 2)

  p1 <-
    ggplot(resdf, aes(x = baseMean, y = log2FoldChange, color = class)) +
    geom_point(stroke = 0,

```

```

        alpha = 0.75,
        size = 0.75,
        show.legend = F) +
xlab('log10 Mean Expression') +
ylab('log2 Fold Change') +
scale_x_log10(
  limits = c(x_lim_left, x_lim_right),
  breaks = breaks_x,
  labels = round(log10(breaks_x)),
  expand = c(0,0)
) +
scale_y_continuous(
  limits = c(y_lim_bottom, y_lim_top),
  expand = c(0,0),
  breaks = breaks_y,
  labels = breaks_y
)+
scale_color_manual(
  values = c("Activated" = "#BB0021",
             "Repressed" = "#3B4992",
             "Unchanged" = "gray"),
)+
geom_hline(yintercept = 0,
           size = 0.5,
           alpha = 1) +
annotate(geom = 'text',
         label = numAct,
         x = min_x,
         y = max_y,
         color = "#BB0021",
         hjust = -0.25,
         vjust = 0.5)+
annotate(geom = 'text',
         label = numRep,
         x = min_x,
         y = min_y,
         color = "#3B4992",
         hjust = -0.25,
         vjust = -0.5)+
ggtheme.jj() +
theme()

if(l2fc_cutoff > 0){
  p1 <- p1+
    geom_hline(yintercept = l2fc_cutoff,
              size = 0.5,
              color = "grey",
              linetype = "dashed")+
  geom_hline(yintercept = -l2fc_cutoff,
            size = 0.5,
            color = "grey",
            linetype = "dashed")
}

```

```

    return(p1)
}

```

Global objects

Scale factors, palettes, color scales, gene list, ATAC/ChIP peaks

```

# Header for reading in DESeq2 results files using read_tsv()
DESeq2_headers.chr <-
  c("name",
    "baseMean",
    "log2FoldChange",
    "lfcSE",
    "stat",
    "pvalue",
    "padj")

# Global palettes and ggplot color scales
RNAi_cols.chr <- c("#BBBBBB", "#66CCEE", "#228833", "#CCBB44", "#AA3377")
updown_cols.chr <- c("#BB0021", "#3B4992")

gg_color_scale_RNAi_ATACDHS <- scale_color_manual(
  values = RNAi_cols.chr,
  name = "RNAi",
  labels = c(
    "LACZ_ATACDHS" = "LACZ",
    "GAF_ATACDHS" = "GAF",
    "BAP170_ATACDHS" = "BAP",
    "NURF301_ATACDHS" = "NURF",
    "NURF301BAP170_ATACDHS" = "NURF+BAP"
  ))

gg_fill_scale_RNAi_ATACDHS <- scale_fill_manual(
  values = RNAi_cols.chr,
  name = "RNAi",
  labels = c(
    "LACZ_ATACDHS" = "LACZ",
    "GAF_ATACDHS" = "GAF",
    "BAP170_ATACDHS" = "BAP",
    "NURF301_ATACDHS" = "NURF",
    "NURF301BAP170_ATACDHS" = "NURF+BAP"
  ))

gg_color_scale_RNAi_ATACMN <- scale_color_manual(
  values = RNAi_cols.chr,
  name = "RNAi",
  labels = c(
    "LACZ_ATACMN" = "LACZ",
    "GAF_ATACMN" = "GAF",
    "BAP170_ATACMN" = "BAP",
    "NURF301_ATACMN" = "NURF",
    "NURF301BAP170_ATACMN" = "NURF+BAP"
  ))

```

```

))

gg_fill_scale_RNAi_ATACMN <- scale_fill_manual(
  values = RNAi_cols.chr,
  name = "RNAi",
  labels = c(
    "LACZ_ATACMN" = "LACZ",
    "GAF_ATACMN" = "GAF",
    "BAP170_ATACMN" = "BAP",
    "NURF301_ATACMN" = "NURF",
    "NURF301BAP170_ATACMN" = "NURF+BAP"
  )
))

gg_color_scale_RNAi_PROseq <- scale_color_manual(
  values = RNAi_cols.chr,
  name = "RNAi",
  labels = c(
    "LACZ_PROseq" = "LACZ",
    "GAF_PROseq" = "GAF",
    "BAP170_PROseq" = "BAP",
    "NURF301_PROseq" = "NURF",
    "NURF301BAP170_PROseq" = "NURF+BAP"
  )
))

gg_fill_scale_RNAi_PROseq <- scale_fill_manual(
  values = RNAi_cols.chr,
  name = "RNAi",
  labels = c(
    "LACZ_PROseq" = "LACZ",
    "GAF_PROseq" = "GAF",
    "BAP170_PROseq" = "BAP",
    "NURF301_PROseq" = "NURF",
    "NURF301BAP170_PROseq" = "NURF+BAP"
  )
))

# PRO-cap corrected gene list
genes.tbl <- read_tsv(
  "genome/dm6_genes_PROcap_corrected.bed",
  col_names = c(
    "chr",
    "start",
    "stop",
    "gene_name",
    "tx_name",
    "strand",
    "pro_cap_shift",
    "pro_cap_signal"
  )
) %>%
  filter(stop - start > 500) # Filter transcripts less than 500 bp long

# PRO-cap TSS correction occasionally condenses two isoforms that have
# closeby TSSs into one.

```

```

# Remove these because they now have duplicate coordinates
# (Brings list from 14818 transcripts to 14474 genes)
genes.tbl <- genes.tbl[which(!duplicated(genes.tbl[, 1:4])), ]

# Making GRanges of genes
genes.gr <-
  tidyChromosomes(
    makeGRangesFromDataFrame(genes.tbl[, -c(7,8)],
                             keep.extra.columns = T))

# ATAC-seq peak summits (called with macs2 using all samples as input)
ATACseq_summits.tbl <- read_tsv(
  "bed/ATACseq_peaks_summits_macs2.bed",
  col_names = c("chr",
                 "start",
                 "stop",
                 "peak_name",
                 "score")
)

ATACseq_summits.gr <-
  tidyChromosomes(
    makeGRangesFromDataFrame(ATACseq_summits.tbl,
                             keep.extra.columns = T))

# GAF ChIP-seq peaks
GAF_peaks.tbl <- read_tsv(
  "bed/GAF_ChIPseq_peaks.narrowPeak",
  col_names = c(
    "chr", "start", "end", "name",
    "displayScore", "strand", "summitFC",
    "mlog10p", "mlog10q", "summitPosition"
  )
)

GAF_peaks.gr <- tidyChromosomes(
  makeGRangesFromDataFrame(GAF_peaks.tbl, keep.extra.columns = T)
)

# PRO-seq scaling factors (derived from human spike-in cells)
# All scale factors are in the format of 1 / x.xxx because DESeq2 divides
# the matrix by the scale factor. These scale factors were designed calculated
# based on scaling by multiplication, so the inverse is provided here for
# consistency and direct compatibility with DESeq2
samples_PROseq.chr <- c(
  "BAP170_PROseq_Rep1",
  "BAP170_PROseq_Rep2",
  "GAF_PROseq_Rep1",
  "GAF_PROseq_Rep2",
  "LACZ_PROseq_Rep1",
  "LACZ_PROseq_Rep2",
  "NURF301BAP170_PROseq_Rep1",
  "NURF301BAP170_PROseq_Rep2",

```

```

    "NURF301_PROseq_Rep1",
    "NURF301_PROseq_Rep2"
)
scale_facts_PROseq.dbl <-      c(
  1 / 0.626244788,
  1 / 1.000000000,
  1 / 0.364059812,
  1 / 0.650773152,
  1 / 0.533494922,
  1 / 0.579464988,
  1 / 0.522119671,
  1 / 0.556802652,
  1 / 0.559650456,
  1 / 0.964053768
)
scale_facts_PROseq.lst <-
  setNames(as.list(scale_facts_PROseq.dbl), samples_PROseq.chr)

# RNA-seq scaling factors (ERCC spike-in derived)
samples_RNAseq.chr <-      c(
  "BAP170_RNAseq_Rep1",
  "BAP170_RNAseq_Rep2",
  "GAF_RNAseq_Rep1",
  "GAF_RNAseq_Rep2",
  "LACZ_RNAseq_Rep1",
  "LACZ_RNAseq_Rep2",
  "NURF301BAP170_RNAseq_Rep1",
  "NURF301BAP170_RNAseq_Rep2",
  "NURF301_RNAseq_Rep1",
  "NURF301_RNAseq_Rep2"
)

scale_facts_RNAseq.dbl <- c(
  1 / 0.696477995,
  1 / 0.645014413,
  1 / 0.75365758,
  1 / 0.596904982,
  1 / 0.744996406,
  1 / 1.00000000,
  1 / 0.707692031,
  1 / 0.490665005,
  1 / 0.668193695,
  1 / 0.772597526
)

scale_facts_RNAseq.lst <-
  setNames(as.list(scale_facts_RNAseq.dbl), samples_RNAseq.chr)

# ATAC-seq DHS (fragments < 120 bp) scaling factors
# (Determined by DESeq2 using counts of DHS signal
# in promoter regions, see below)
samples_ATACDHS.chr <-      c(
  "BAP170_ATACDHS_Rep1",

```

```

    "BAP170_ATACDHS_Rep2",
    "GAF_ATACDHS_Rep1",
    "GAF_ATACDHS_Rep2",
    "LACZ_ATACDHS_Rep1",
    "LACZ_ATACDHS_Rep2",
    "NURF301BAP170_ATACDHS_Rep1",
    "NURF301BAP170_ATACDHS_Rep2",
    "NURF301_ATACDHS_Rep1",
    "NURF301_ATACDHS_Rep2"
)

scale_facts_ATACDHS.dbl <- c(
  1 / 1,
  1 / 1,
  1 / 1,
  1 / 1,
  1 / 1,
  1 / 1,
  1 / 1,
  1 / 1,
  1 / 1,
  1 / 1,
  1 / 1
)

scale_facts_ATACDHS.lst <-
  setNames(as.list(scale_facts_ATACDHS.dbl), samples_ATACDHS.chr)

# ATAC-seq MN (mononucs, 130 < fragment < 200) scaling factors
# (Determined by DESeq2 using counts of DHS signal
# in promoter regions, see below)
samples_ATACMN.chr <- c(
  "BAP170_ATACMN_Rep1",
  "BAP170_ATACMN_Rep2",
  "GAF_ATACMN_Rep1",
  "GAF_ATACMN_Rep2",
  "LACZ_ATACMN_Rep1",
  "LACZ_ATACMN_Rep2",
  "NURF301BAP170_ATACMN_Rep1",
  "NURF301BAP170_ATACMN_Rep2",
  "NURF301_ATACMN_Rep1",
  "NURF301_ATACMN_Rep2"
)

scale_facts_ATACMN.dbl <- c(
  1 / 1,
  1 / 1,
  1 / 1,
  1 / 1,
  1 / 1,
  1 / 1,
  1 / 1,
  1 / 1,
  1 / 1,
  1 / 1,
  1 / 1,
  1 / 1
)

```

```

1 / 1
)

scale_facts_ATACMN.lst <-
  setNames(as.list(scale_facts_ATACMN.dbl), samples_ATACMN.chr)

```

Reading in PROseq bw files

```

PROseq.lst <-
  list(
    "BAP170_PROseq_Rep1" = import_bigWig(
      "bw/PROseq/raw_signal/BAP170_PROseq_Rep1_fwd.bw",
      "bw/PROseq/raw_signal/BAP170_PROseq_Rep1_rev.bw"
    ),
    "BAP170_PROseq_Rep2" = import_bigWig(
      "bw/PROseq/raw_signal/BAP170_PROseq_Rep2_fwd.bw",
      "bw/PROseq/raw_signal/BAP170_PROseq_Rep2_rev.bw"
    ),
    "GAF_PROseq_Rep1" = import_bigWig(
      "bw/PROseq/raw_signal/GAF_PROseq_Rep1_fwd.bw",
      "bw/PROseq/raw_signal/GAF_PROseq_Rep1_rev.bw"
    ),
    "GAF_PROseq_Rep2" = import_bigWig(
      "bw/PROseq/raw_signal/GAF_PROseq_Rep2_fwd.bw",
      "bw/PROseq/raw_signal/GAF_PROseq_Rep2_rev.bw"
    ),
    "LACZ_PROseq_Rep1" = import_bigWig(
      "bw/PROseq/raw_signal/LACZ_PROseq_Rep1_fwd.bw",
      "bw/PROseq/raw_signal/LACZ_PROseq_Rep1_rev.bw"
    ),
    "LACZ_PROseq_Rep2" = import_bigWig(
      "bw/PROseq/raw_signal/LACZ_PROseq_Rep2_fwd.bw",
      "bw/PROseq/raw_signal/LACZ_PROseq_Rep2_rev.bw"
    ),
    "NURF301BAP170_PROseq_Rep1" = import_bigWig(
      "bw/PROseq/raw_signal/NURF301BAP170_PROseq_Rep1_fwd.bw",
      "bw/PROseq/raw_signal/NURF301BAP170_PROseq_Rep1_rev.bw"
    ),
    "NURF301BAP170_PROseq_Rep2" = import_bigWig(
      "bw/PROseq/raw_signal/NURF301BAP170_PROseq_Rep2_fwd.bw",
      "bw/PROseq/raw_signal/NURF301BAP170_PROseq_Rep2_rev.bw"
    ),
    "NURF301_PROseq_Rep1" = import_bigWig(
      "bw/PROseq/raw_signal/NURF301_PROseq_Rep1_fwd.bw",
      "bw/PROseq/raw_signal/NURF301_PROseq_Rep1_rev.bw"
    ),
    "NURF301_PROseq_Rep2" = import_bigWig(
      "bw/PROseq/raw_signal/NURF301_PROseq_Rep2_fwd.bw",
      "bw/PROseq/raw_signal/NURF301_PROseq_Rep2_rev.bw"
    )
  )

```


Reading in RNAseq bw files

```
RNAseq.lst <-  
list(  
  "BAP170_RNAseq_Rep1" = import_bigWig(  
    "bw/RNAseq/raw_signal/BAP_RNAseq_Rep1_fwd.bw",  
    "bw/RNAseq/raw_signal/BAP_RNAseq_Rep1_rev.bw"  
  ),  
  "BAP170_RNAseq_Rep2" = import_bigWig(  
    "bw/RNAseq/raw_signal/BAP_RNAseq_Rep2_fwd.bw",  
    "bw/RNAseq/raw_signal/BAP_RNAseq_Rep2_rev.bw"  
  ),  
  "GAF_RNAseq_Rep1" = import_bigWig(  
    "bw/RNAseq/raw_signal/GAF_RNAseq_Rep1_fwd.bw",  
    "bw/RNAseq/raw_signal/GAF_RNAseq_Rep1_rev.bw"  
  ),  
  "GAF_RNAseq_Rep2" = import_bigWig(  
    "bw/RNAseq/raw_signal/GAF_RNAseq_Rep2_fwd.bw",  
    "bw/RNAseq/raw_signal/GAF_RNAseq_Rep2_rev.bw"  
  ),  
  "LACZ_RNAseq_Rep1" = import_bigWig(  
    "bw/RNAseq/raw_signal/LACZ_RNAseq_Rep1_fwd.bw",  
    "bw/RNAseq/raw_signal/LACZ_RNAseq_Rep1_rev.bw"  
  ),  
  "LACZ_RNAseq_Rep2" = import_bigWig(  
    "bw/RNAseq/raw_signal/LACZ_RNAseq_Rep2_fwd.bw",  
    "bw/RNAseq/raw_signal/LACZ_RNAseq_Rep2_rev.bw"  
  ),  
  "NURF301_RNAseq_Rep1" = import_bigWig(  
    "bw/RNAseq/raw_signal/NURF_RNAseq_Rep1_fwd.bw",  
    "bw/RNAseq/raw_signal/NURF_RNAseq_Rep1_rev.bw"  
  ),  
  "NURF301_RNAseq_Rep2" = import_bigWig(  
    "bw/RNAseq/raw_signal/NURF_RNAseq_Rep2_fwd.bw",  
    "bw/RNAseq/raw_signal/NURF_RNAseq_Rep2_rev.bw"  
  ),  
  "NURF301BAP170_RNAseq_Rep1" = import_bigWig(  
    "bw/RNAseq/raw_signal/NURFBAP_RNAseq_Rep1_fwd.bw",  
    "bw/RNAseq/raw_signal/NURFBAP_RNAseq_Rep1_rev.bw"  
  ),  
  "NURF301BAP170_RNAseq_Rep2" = import_bigWig(  
    "bw/RNAseq/raw_signal/NURFBAP_RNAseq_Rep2_fwd.bw",  
    "bw/RNAseq/raw_signal/NURFBAP_RNAseq_Rep2_rev.bw"  
  )  
)
```

Reading in ATAC-seq bw files

```
# First DHS, which is ATAC-seq that's been filtered for fragments < 120 bp  
# only (entire fragments piled up for signal)  
ATACDHS.lst <-
```

```

list(
  "BAP170_ATACDHS_Rep1" =
    tidyChromosomes(
      import.bw("bw/ATACseq/DHS/raw_signal/BAP170_Rep1_merged_AtacDHS.bw")
    ),
  "BAP170_ATACDHS_Rep2" =
    tidyChromosomes(
      import.bw("bw/ATACseq/DHS/raw_signal/BAP170_Rep2_merged_AtacDHS.bw")
    ),
  "GAF_ATACDHS_Rep1" =
    tidyChromosomes(
      import.bw("bw/ATACseq/DHS/raw_signal/GAF_Rep1_merged_AtacDHS.bw")
    ),
  "GAF_ATACDHS_Rep2" =
    tidyChromosomes(
      import.bw("bw/ATACseq/DHS/raw_signal/GAF_Rep2_merged_AtacDHS.bw")
    ),
  "LACZ_ATACDHS_Rep1" =
    tidyChromosomes(
      import.bw("bw/ATACseq/DHS/raw_signal/LACZ_Rep1_merged_AtacDHS.bw")
    ),
  "LACZ_ATACDHS_Rep2" =
    tidyChromosomes(
      import.bw("bw/ATACseq/DHS/raw_signal/LACZ_Rep2_merged_AtacDHS.bw")
    ),
  "NURF301_ATACDHS_Rep1" =
    tidyChromosomes(
      import.bw("bw/ATACseq/DHS/raw_signal/NURF301_Rep1_merged_AtacDHS.bw")
    ),
  "NURF301_ATACDHS_Rep2" =
    tidyChromosomes(
      import.bw("bw/ATACseq/DHS/raw_signal/NURF301_Rep2_merged_AtacDHS.bw")
    ),
  "NURF301BAP170_ATACDHS_Rep1" =
    tidyChromosomes(
      import.bw(
        "bw/ATACseq/DHS/raw_signal/NURF301BAP170_Rep1_merged_AtacDHS.bw"
      )
    ),
  "NURF301BAP170_ATACDHS_Rep2" =
    tidyChromosomes(
      import.bw(
        "bw/ATACseq/DHS/raw_signal/NURF301BAP170_Rep2_merged_AtacDHS.bw"
      )
    )
)

# Now MN, which is ATAC-seq thats 130 bp < fragment < 200 bp
# (only central 3 bp contribute to signal)

ATACMN.lst <-
list(
  "BAP170_ATACMN_Rep1" =

```

```

    tidyChromosomes(
      import.bw("bw/ATACseq/MN/raw_signal/BAP170_Rep1_merged_AtacMN.bw")
    ),
    "BAP170_ATACMN_Rep2" =
      tidyChromosomes(
        import.bw("bw/ATACseq/MN/raw_signal/BAP170_Rep2_merged_AtacMN.bw")
      ),
    "GAF_ATACMN_Rep1" =
      tidyChromosomes(
        import.bw("bw/ATACseq/MN/raw_signal/GAF_Rep1_merged_AtacMN.bw")
      ),
    "GAF_ATACMN_Rep2" =
      tidyChromosomes(
        import.bw("bw/ATACseq/MN/raw_signal/GAF_Rep2_merged_AtacMN.bw")
      ),
    "LACZ_ATACMN_Rep1" =
      tidyChromosomes(
        import.bw("bw/ATACseq/MN/raw_signal/LACZ_Rep1_merged_AtacMN.bw")
      ),
    "LACZ_ATACMN_Rep2" =
      tidyChromosomes(
        import.bw("bw/ATACseq/MN/raw_signal/LACZ_Rep2_merged_AtacMN.bw")
      ),
    "NURF301_ATACMN_Rep1" =
      tidyChromosomes(
        import.bw("bw/ATACseq/MN/raw_signal/NURF301_Rep1_merged_AtacMN.bw")
      ),
    "NURF301_ATACMN_Rep2" =
      tidyChromosomes(
        import.bw("bw/ATACseq/MN/raw_signal/NURF301_Rep2_merged_AtacMN.bw")
      ),
    "NURF301BAP170_ATACMN_Rep1" =
      tidyChromosomes(
        import.bw(
          "bw/ATACseq/MN/raw_signal/NURF301BAP170_Rep1_merged_AtacMN.bw"
        )
      ),
    "NURF301BAP170_ATACMN_Rep2" =
      tidyChromosomes(
        import.bw(
          "bw/ATACseq/MN/raw_signal/NURF301BAP170_Rep2_merged_AtacMN.bw"
        )
      )
  )
)

```

Reading in ChIP/CUT&RUN bw files

```

# Reading in list of CUT&RUN bigWigs
CaR.lst <-
list(
  "GAF_CUTandRUN" = tidyChromosomes(
    import.bw("bw/CutAndRun/GAF_CUTandRUN.bw")
  )
)

```

```

    ),
    "NURF_CUTandRUN" = tidyChromosomes(
      import.bw("bw/CutAndRun/NURF_CUTandRUN.bw")
    )
  )

# RPM normalizing CUT&RUN signal
CaR.lst <- lapply(CaR.lst, RPMnorm)

# Reading in list of ChIP-seq bigWigs (published data)
ChIP.lst <-
  list(
    "GAF_ChIPseq" = tidyChromosomes(
      import.bw("bw/ChIPseq/GAF_ChIPseq.bw")
    ),
    "M1BP_ChIPseq" = tidyChromosomes(
      import.bw("bw/ChIPseq/M1BP_ChIPseq.bw")
    ),
    "BEAF32_ChIPseq" = tidyChromosomes(
      import.bw("bw/ChIPseq/BEAF32_ChIPseq.bw")
    )
  )

# RPM normalizing ChIP-seq signal
ChIP.lst <- lapply(ChIP.lst, RPMnorm)

# Reading in list of new ChIP-seq (generated for this paper)
ChIPnew.lst <-
  list(
    "LACZRNAi_GAF_ChIPseq_Rep1" = tidyChromosomes(
      import.bw("bw/ChIPseq/raw_signal/LACZ_GAF_Rep1.bw")
    ),
    "LACZRNAi_GAF_ChIPseq_Rep2" = tidyChromosomes(
      import.bw("bw/ChIPseq/raw_signal/LACZ_GAF_Rep2.bw")
    ),
    "GAFRNAi_GAF_ChIPseq_Rep1" = tidyChromosomes(
      import.bw("bw/ChIPseq/raw_signal/GAF_GAF_Rep1.bw")
    ),
    "GAFRNAi_GAF_ChIPseq_Rep2" = tidyChromosomes(
      import.bw("bw/ChIPseq/raw_signal/GAF_GAF_Rep2.bw")
    ),
    "BAPRNAi_GAF_ChIPseq_Rep1" = tidyChromosomes(
      import.bw("bw/ChIPseq/raw_signal/BAP_GAF_Rep1.bw")
    ),
    "BAPRNAi_GAF_ChIPseq_Rep2" = tidyChromosomes(
      import.bw("bw/ChIPseq/raw_signal/BAP_GAF_Rep2.bw")
    )
  )

ChIPnew_merged.lst <- list(
  "LACZRNAi_GAF_ChIPseq" = mergeGRangesData(
    ChIPnew.lst$LACZRNAi_GAF_ChIPseq_Rep1,

```

```

    CHIPnew.lst$LACZRNAi_GAF_ChIPseq_Rep2
  ),
  "GAFRNAi_GAF_ChIPseq" = mergeGRangesData(
    CHIPnew.lst$GAFRNAi_GAF_ChIPseq_Rep1,
    CHIPnew.lst$GAFRNAi_GAF_ChIPseq_Rep2
  ),
  "BAPRNAi_GAF_ChIPseq" = mergeGRangesData(
    CHIPnew.lst$BAPRNAi_GAF_ChIPseq_Rep1,
    CHIPnew.lst$BAPRNAi_GAF_ChIPseq_Rep2
  )
)

# RPM normalizing new ChIP-seq data
CHIPnew_merged.lst <- lapply(CHIPnew_merged.lst, RPMnorm)

```

Generating count tables

```

# Gene body (TSS+200 to TES-200)
# initialize matrix
GB.tbl <- tibble("tx_name" = genes.tbl$tx_name)
# populate with counts
for(i in samples_PROseq.chr){
  GB.tbl[i] <- getCountsByRegions(
    PROseq.lst[[i]], genebodies(genes.gr, 200, -200))
}

## Promoter (TSS-50 to TSS+100)
PR.tbl <- tibble("tx_name" = genes.tbl$tx_name)
for(i in samples_PROseq.chr){
  PR.tbl[i] <- getCountsByRegions(
    PROseq.lst[[i]], genebodies(genes.gr, -50, 100, fix.end = "start"))
}

## Upstream (TSS-500 to TSS-200)
US.tbl <- tibble("tx_name" = genes.tbl$tx_name)
for(i in samples_PROseq.chr){
  US.tbl[i] <- getCountsByRegions(
    PROseq.lst[[i]], genebodies(genes.gr, -500, -200, fix.end = "start"))
}

# Getting RNA-seq count data
RNA.tbl <- tibble("tx_name" = genes.tbl$tx_name)
for(i in samples_RNAseq.chr){
  RNA.tbl[i] <- getCountsByRegions(
    RNAseq.lst[[i]], genebodies(genes.gr, -1000, 0, fix.start = "end"))
}

# Getting ATAC-seq counts in peaks
ATAC_peaks.tbl <- tibble("peak_name" = ATACseq_summits.gr$peak_name)
for(i in samples_ATACDHS.chr){

```

```

ATAC_peaks.tbl[i] <-
  getCountsByRegions(
    ATACDHS.lst[[i]],
    promoters(ATACseq_summits.gr, 100, 100),
    expand_ranges = TRUE)
}

# Getting ATAC-seq counts at promoters
ATAC_PR.tbl <- tibble("tx_name" = genes.gr$tx_name)
for(i in samples_ATACDHS.chr){
  ATAC_PR.tbl[i] <-
    getCountsByRegions(
      ATACDHS.lst[[i]],
      promoters(genes.gr, 1000, 0),
      expand_ranges = TRUE)
}

# Getting ATAC-seq counts (mononucleosome sized fragments)
# in the first 1.5kb of each gene
ATACMN_genes.tbl <- tibble("tx_name" = genes.gr$tx_name)
for(i in samples_ATACMN.chr){
  ATACMN_genes.tbl[i] <-
    getCountsByRegions(
      ATACMN.lst[[i]], promoters(genes.gr, 0, 1500), expand_ranges = TRUE)
}

```

Blacklisting genes with too much upstream transcription

If an upstream gene is differentially expressed in one condition, it can appear like a downstream gene has differentially expressed pausing due to read-through transcription (see Duarte et al. Genes Dev 2016. doi:10.1101/gad.284430.116 for more detailed explanation). To eliminate this problem, we filter out any genes in our gene list that have more than half the PRO-seq signal in the upstream region (TSS-500 to TSS-200) as is observed in the pause region (TSS-50 to TSS+100), or genes which have more signal in the upstream region than in the length normalized gene body. This results in blacklisting 5099 transcripts, or ~35% of annotated unique transcripts over 500 bp, leaving 9375 genes for further analysis

```

# Normalizing data using spike-in scale factors
# Gene body (TSS+200 to TES-200)
GB_normed.tbl <- GB.tbl
for (i in colnames(GB.tbl[, -1])) {
  GB_normed.tbl[, i] <- GB.tbl[, i] / scale_facts_PROseq.lst[[i]]
}

# Promoter (TSS-50 to TSS+100)
PR_normed.tbl <- PR.tbl
for (i in colnames(PR.tbl[, -1])) {
  PR_normed.tbl[, i] <- PR.tbl[, i] / scale_facts_PROseq.lst[[i]]
}

# Upstream (TSS-500 to TSS-200)
US_normed.tbl <- US.tbl
for (i in colnames(US.tbl[, -1])) {
  US_normed.tbl[, i] <- US.tbl[, i] / scale_facts_PROseq.lst[[i]]
}

```

```

}

# Initializing data frame of replicate averages of
# LACZ for PR, GB, & US for each transcript. Also
# getting gene length to length matched normalize
# GB signal
blacklist.tbl <- data.frame(
  "tx_name" = GB_normed.tbl$tx_name,
  "gene_length" = genes.tbl$stop - genes.tbl$start,
  "GB" = apply(
    GB_normed.tbl[, c("LACZ_PROseq_Rep1", "LACZ_PROseq_Rep2")], 1, mean
  ),
  "PR" = apply(
    PR_normed.tbl[, c("LACZ_PROseq_Rep1", "LACZ_PROseq_Rep2")], 1, mean
  ),
  "US" = apply(
    US_normed.tbl[, c("LACZ_PROseq_Rep1", "LACZ_PROseq_Rep2")], 1, mean
  )
)

# Length normalizing GB to RPK
blacklist.tbl$GB <-
  blacklist.tbl$GB * (1000 / blacklist.tbl$gene_length)

# Length normalizing PR to RPK
blacklist.tbl$PR <-
  blacklist.tbl$PR * (1000 / 150)

# Length normalizing US to RPK
blacklist.tbl$US <-
  blacklist.tbl$US * (1000 / 300)

# Blacklisting genes with US transcription equal to more than 0.5X PR or GB
blacklist.chr <- (unique(rbind(
  blacklist.tbl[which(blacklist.tbl$US > (0.5 * blacklist.tbl$PR)),],
  blacklist.tbl[which(blacklist.tbl$US > blacklist.tbl$GB),])
))$tx_name

# Filtering blacklisted genes out of all data loaded so far
GB.tbl <-
  GB.tbl[which(!(GB.tbl$tx_name %in% blacklist.chr)),]
PR.tbl <-
  PR.tbl[which(!(PR.tbl$tx_name %in% blacklist.chr)),]
GB_normed.tbl <-
  GB_normed.tbl[which(!(GB_normed.tbl$tx_name %in% blacklist.chr)),]
PR_normed.tbl <-
  PR_normed.tbl[which(!(PR_normed.tbl$tx_name %in% blacklist.chr)),]
genes.tbl <-
  genes.tbl[which(!(genes.tbl$tx_name %in% blacklist.chr)),]
genes.gr <-
  genes.gr[which(!(genes.gr$tx_name %in% blacklist.chr)),]
ATAC_PR.tbl <-

```

```

    ATAC_PR.tbl[which(!(ATAC_PR.tbl$tx_name %in% blacklist.chr)),]
    ATACMN_genes.tbl <-
    ATACMN_genes.tbl[which(!(ATACMN_genes.tbl$tx_name %in% blacklist.chr)),]

# Saving gene list for loading later
write_tsv(genes.tbl[,c(7, 8)], path = "bed/filtered_dm6_genes.bed", col_names = FALSE)

```

Normalizing, pooling replicates, and saving bw files

```

# Normalizing PRO-seq in place using scale factors
for (i in samples_PROseq.chr) {
    PROseq.lst[[i]]$score <- PROseq.lst[[i]]$score / scale_facts_PROseq.lst[[i]]
}

# Merging replicates (summing them across bins) and saving
PROseq_normed_merged.lst <- vector(mode = "list", length = OL)
for(i in samples_PROseq.chr[
    which(unlist(lapply(
        samples_PROseq.chr, grepl, pattern = "Rep1"
    )))]
) {
    PROseq_normed_merged.lst[[gsub("_Rep1", "", i)]] <-
        mergeRangesData(
            PROseq.lst[[i]],
            PROseq.lst[[gsub("_Rep1", "_Rep2", i)]]
        )
}

# Saving bw files for future loading
for(i in names(PROseq_normed_merged.lst)){
    export.bw(
        PROseq_normed_merged.lst[[i]][which(
            strand(PROseq_normed_merged.lst[[i]]) == "+"
        )],
        paste0("bw/PROseq/merged_normed/", gsub("seq", "seq_fwd.bw", i))
    )
    export.bw(
        PROseq_normed_merged.lst[[i]][which(
            strand(PROseq_normed_merged.lst[[i]]) == "-"
        )],
        paste0("bw/PROseq/merged_normed/", gsub("seq", "seq_rev.bw", i))
    )
}

# Normalizing RNA-seq in place using scale factors
for (i in samples_RNAseq.chr) {
    RNAseq.lst[[i]]$score <-
    RNAseq.lst[[i]]$score / scale_facts_RNAseq.lst[[i]]
}

# Merging replicates (summing them across bins) and saving
RNAseq_normed_merged.lst <- vector(mode = "list", length = OL)
for(i in samples_RNAseq.chr[

```



```

which(unlist(lapply(
  samples_RNAseq.chr, grepl, pattern = "Rep1"
)))]
) {
  RNAseq_normed_merged.lst[[gsub("_Rep1", "", i)]] <-
    mergeGRangesData(
      RNAseq.lst[[i]],
      RNAseq.lst[[gsub("_Rep1", "_Rep2", i)]]
    )
}

# Saving bw files for future loading
for(i in names(RNAseq_normed_merged.lst)){
  export.bw(
    RNAseq_normed_merged.lst[[i]][which(
      strand(RNAseq_normed_merged.lst[[i]]) == "+" )],
    paste0("bw/RNAseq/merged_normed/", gsub("seq", "seq_fwd.bw", i))
  )
  export.bw(
    RNAseq_normed_merged.lst[[i]][which(
      strand(RNAseq_normed_merged.lst[[i]]) == "-" )],
    paste0("bw/RNAseq/merged_normed/", gsub("seq", "seq_rev.bw", i))
  )
}

# Normalizing ATACDHS in place using scale factors
for (i in samples_ATACDHS.chr) {
  ATACDHS.lst[[i]]$score <-
    ATACDHS.lst[[i]]$score / scale_facts_ATACDHS.lst[[i]]
}

# Merging replicates (summing them across bins) and saving
BAP170_merged.gr <-
  mergeGRangesData(
    ATACDHS.lst["BAP170_ATACDHS_Rep1"],
    ATACDHS.lst["BAP170_ATACDHS_Rep2"])
export.bw(
  BAP170_merged.gr,
  "bw/ATACseq/DHS/merged_normed/BAP170_ATACDHS.bw")

GAF_merged.gr <-
  mergeGRangesData(
    ATACDHS.lst["GAF_ATACDHS_Rep1"],
    ATACDHS.lst["GAF_ATACDHS_Rep2"])
export.bw(
  GAF_merged.gr,
  "bw/ATACseq/DHS/merged_normed/GAF_ATACDHS.bw")

LACZ_merged.gr <-
  mergeGRangesData(
    ATACDHS.lst["LACZ_ATACDHS_Rep1"],
    ATACDHS.lst["LACZ_ATACDHS_Rep2"])
export.bw(

```

```

LACZ_merged.gr,
"bw/ATACseq/DHS/merged_normed/LACZ_ATACDHS.bw")

NURF301_merged.gr <-
  mergeGRangesData(
    ATACDHS.lst["NURF301_ATACDHS_Rep1"],
    ATACDHS.lst["NURF301_ATACDHS_Rep2"])
export.bw(
  NURF301_merged.gr,
  "bw/ATACseq/DHS/merged_normed/NURF301_ATACDHS.bw")

NURF301BAP170_merged.gr <-
  mergeGRangesData(
    ATACDHS.lst["NURF301BAP170_ATACDHS_Rep1"],
    ATACDHS.lst["NURF301BAP170_ATACDHS_Rep2"])
export.bw(
  NURF301BAP170_merged.gr,
  "bw/ATACseq/DHS/merged_normed/NURF301BAP170_ATACDHS.bw")

# Getting DESeq Normalization factors for ATACMN data
# Getting counts matrix
ATACMN_genes.df <-
  data.frame(ATACMN_genes.tbl, row.names = "tx_name")

# colData object for DESeq2
col_data.df <- data.frame(
  row.names = colnames(ATACMN_genes.df),
  RNAi = c(
    "BAP170",
    "BAP170",
    "GAF",
    "GAF",
    "LACZ",
    "LACZ",
    "NURF301BAP170",
    "NURF301BAP170",
    "NURF301",
    "NURF301"
  )
)

# construction of DESeq2 object
ATACMN_genes.dds <-
  DESeqDataSetFromMatrix(countData = ATACMN_genes.df,
    colData = col_data.df,
    design = ~ RNAi)

# Running DESeq2
ATACMN_genes.dds <- DESeq(ATACMN_genes.dds)

scale_facts_ATACMN.db1 <- sizeFactors(ATACMN_genes.dds)

```

```

scale_facts_ATACMN.lst <-
  setNames(as.list(scale_facts_ATACMN.dbl), samples_ATACMN.chr)

# Normalizing ATACMN in place using scale factors
for (i in samples_ATACMN.chr) {
  ATACMN.lst[[i]]$score <-
    ATACMN.lst[[i]]$score / scale_facts_ATACMN.lst[[i]]
}

# Merging replicates (summing them across bins) and saving
BAP170_merged.gr <-
  mergeGRangesData(
    ATACMN.lst["BAP170_ATACMN_Rep1"],
    ATACMN.lst["BAP170_ATACMN_Rep2"])
export.bw(
  BAP170_merged.gr,
  "bw/ATACseq/MN/merged_normed/BAP170_ATACMN.bw")

GAF_merged.gr <-
  mergeGRangesData(
    ATACMN.lst["GAF_ATACMN_Rep1"],
    ATACMN.lst["GAF_ATACMN_Rep2"])
export.bw(
  GAF_merged.gr,
  "bw/ATACseq/MN/merged_normed/GAF_ATACMN.bw")

LACZ_merged.gr <-
  mergeGRangesData(
    ATACMN.lst["LACZ_ATACMN_Rep1"],
    ATACMN.lst["LACZ_ATACMN_Rep2"])
export.bw(
  LACZ_merged.gr,
  "bw/ATACseq/MN/merged_normed/LACZ_ATACMN.bw")

NURF301_merged.gr <-
  mergeGRangesData(
    ATACMN.lst["NURF301_ATACMN_Rep1"],
    ATACMN.lst["NURF301_ATACMN_Rep2"])
export.bw(
  NURF301_merged.gr,
  "bw/ATACseq/MN/merged_normed/NURF301_ATACMN.bw")

NURF301BAP170_merged.gr <-
  mergeGRangesData(
    ATACMN.lst["NURF301BAP170_ATACMN_Rep1"],
    ATACMN.lst["NURF301BAP170_ATACMN_Rep2"])
export.bw(
  NURF301BAP170_merged.gr,
  "bw/ATACseq/MN/merged_normed/NURF301BAP170_ATACMN.bw")

# Merging and saving new merged normalized ChIP-seq
export.bw(
  ChIPnew_merged.lst$LACZRNAi_GAF_ChIPseq,

```

```

"bw/ChIPseq/merged_normalized/LACZRNAi_GAFChIP.bw"
)

export.bw(
  ChIPnew_merged.lst$GAFRNAi_GAF_ChIPseq,
  "bw/ChIPseq/merged_normalized/GAFRNAi_GAFChIP.bw"
)

export.bw(
  ChIPnew_merged.lst$BAPRNAi_GAF_ChIPseq,
  "bw/ChIPseq/merged_normalized/BAPRNAi_GAFChIP.bw"
)

```

Reading back in normalized data

```

rm(ATACDHS.lst)
rm(PROseq.lst)
rm(RNAseq.lst)
rm(ATACMN.lst)

# ATACMN
ATACMN_normed_merged.lst <-
  list(
    "LACZ_ATACMN" =
      tidyChromosomes(
        import.bw("bw/ATACseq/MN/merged_normed/LACZ_ATACMN.bw")
      ),
    "GAF_ATACMN" =
      tidyChromosomes(
        import.bw("bw/ATACseq/MN/merged_normed/GAF_ATACMN.bw")
      ),
    "BAP170_ATACMN" =
      tidyChromosomes(
        import.bw("bw/ATACseq/MN/merged_normed/BAP170_ATACMN.bw")
      ),
    "NURF301_ATACMN" =
      tidyChromosomes(
        import.bw("bw/ATACseq/MN/merged_normed/NURF301_ATACMN.bw")
      ),
    "NURF301BAP170_ATACMN" =
      tidyChromosomes(
        import.bw("bw/ATACseq/MN/merged_normed/NURF301BAP170_ATACMN.bw")
      )
  )

# ATACDHS
ATACDHS_normed_merged.lst <-
  list(
    "LACZ_ATACDHS" =
      tidyChromosomes(
        import.bw("bw/ATACseq/DHS/merged_normed/LACZ_ATACDHS.bw")
      )
  )

```

```

    ),
"GAF_ATACDHS" =
  tidyChromosomes(
    import.bw("bw/ATACseq/DHS/merged_normed/GAF_ATACDHS.bw")
  ),
"BAP170_ATACDHS" =
  tidyChromosomes(
    import.bw("bw/ATACseq/DHS/merged_normed/BAP170_ATACDHS.bw")
  ),
"NURF301_ATACDHS" =
  tidyChromosomes(
    import.bw("bw/ATACseq/DHS/merged_normed/NURF301_ATACDHS.bw")),
"NURF301BAP170_ATACDHS" =
  tidyChromosomes(
    import.bw("bw/ATACseq/DHS/merged_normed/NURF301BAP170_ATACDHS.bw")
  )
)

# PROseq
PROseq_normed_merged.lst <-
  list(
    "LACZ_PROseq" = import_bigWig(
      "bw/PROseq/merged_normed/LACZ_PROseq_fwd.bw",
      "bw/PROseq/merged_normed/LACZ_PROseq_rev.bw"
    ),
    "GAF_PROseq" = import_bigWig(
      "bw/PROseq/merged_normed/GAF_PROseq_fwd.bw",
      "bw/PROseq/merged_normed/GAF_PROseq_rev.bw"
    ),
    "BAP170_PROseq" = import_bigWig(
      "bw/PROseq/merged_normed/BAP170_PROseq_fwd.bw",
      "bw/PROseq/merged_normed/BAP170_PROseq_rev.bw"
    ),
    "NURF301_PROseq" = import_bigWig(
      "bw/PROseq/merged_normed/NURF301_PROseq_fwd.bw",
      "bw/PROseq/merged_normed/NURF301_PROseq_rev.bw"
    ),
    "NURF301BAP170_PROseq" = import_bigWig(
      "bw/PROseq/merged_normed/NURF301BAP170_PROseq_fwd.bw",
      "bw/PROseq/merged_normed/NURF301BAP170_PROseq_rev.bw"
    )
  )

# RNaseq
RNaseq_normed_merged.lst<-
  list(
    "LACZ_RNaseq" = import_bigWig(
      "bw/RNaseq/merged_normed/LACZ_RNaseq_fwd.bw",
      "bw/RNaseq/merged_normed/LACZ_RNaseq_rev.bw"
    ),
    "GAF_RNaseq" = import_bigWig(
      "bw/RNaseq/merged_normed/GAF_RNaseq_fwd.bw",
      "bw/RNaseq/merged_normed/GAF_RNaseq_rev.bw"
    )
  )

```

```

    ),
    "BAP170_RNAseq" = import_bigWig(
      "bw/RNAseq/merged_normed/BAP170_RNAseq_fwd.bw",
      "bw/RNAseq/merged_normed/BAP170_RNAseq_rev.bw"
    ),
    "NURF301_RNAseq" = import_bigWig(
      "bw/RNAseq/merged_normed/NURF301_RNAseq_fwd.bw",
      "bw/RNAseq/merged_normed/NURF301_RNAseq_rev.bw"
    ),
    "NURF301BAP170_RNAseq" = import_bigWig(
      "bw/RNAseq/merged_normed/NURF301BAP170_RNAseq_fwd.bw",
      "bw/RNAseq/merged_normed/NURF301BAP170_RNAseq_rev.bw"
    )
  )

genes.gr <- makeGRangesFromDataFrame(
  read_tsv("bed/filtered_dm6_genes.bed",
    col_names = c(
      "chr", "start", "end",
      "gene_name", "tx_name",
      "strand"
    )),
  keep.extra.columns = T
)

# Reading in M1BP PRO-seq
PROseq_M1BP.lst <- list(
  "LACZ" = import_bigWig(
    "bw/M1BP_PROseq/LACZ_fwd.bw",
    "bw/M1BP_PROseq/LACZ_rev.bw"
  ),
  "M1BP" = import_bigWig(
    "bw/M1BP_PROseq/M1BP_fwd.bw",
    "bw/M1BP_PROseq/M1BP_rev.bw"
  )
)

```

Figure 1

A - Overview cartoon (not made in R, inserted in Illustrator later)
 B - PCA of promoter PRO-seq
 C - glob1 browser shot D - MA Plot of Promoter region PRO-seq GAF vs. LACZ E - GAF-dependent promoters ATAC-seq (<120bp) metaprofile
 F - GAF-dependent promoters PRO-seq (<120bp) metaprofile
 G - GAF ChIP-seq heatmaps in different RNAi conditions H - Duplicate of E - easier to remove irrelevant lines from each later
 I - GAF PR l2FC vs PBAP and NURF scatterplot

```

# Panel A - just use plot spacer and insert cartoon in Illustrator later

# Panel B - PCA of promoter PRO-seq
# Formatting count matrix for DESeq2.
# Rownames = gene names, only count data

```

```

PR.df <- data.frame(PR.tbl, row.names = "tx_name")

# colData object for DESeq2
col_data.df <- data.frame(row.names = colnames(PR.df),
  RNAi = c("BAP170", "BAP170",
    "GAF", "GAF",
    "LACZ", "LACZ",
    "NURF301BAP170", "NURF301BAP170",
    "NURF301", "NURF301"))

# Construction of DESeq2 object
PR.dds <- DESeqDataSetFromMatrix(countData = PR.df,
  colData = col_data.df,
  design = ~ RNAi)

# Adding spike-in scale factors
sizeFactors(PR.dds) <- scale_facts_PROseq.dbl

# Running DESeq2
PR.dds <- DESeq(PR.dds)

# Variance stabilizing log transform
PR.rld <- rlog(PR.dds, blind = F)

# Getting PCA data
PR.pca <-
  plotPCA(PR.rld,
    intgroup = c("RNAi"),
    returnData = TRUE)
# Getting % Var explained by each PC
PR_pcavar.dbl <- round(100 * attr(PR.pca, "percentVar"))

# Factoring for plotting
PR.pca$RNAi <-
  factor(PR.pca$RNAi,
    levels = c("LACZ", "GAF", "BAP170", "NURF301", "NURF301BAP170"))

# Plotting
F1B <- ggplot(PR.pca, aes(PC1, PC2, color = RNAi)) +
  geom_point(size = 1.5) +
  xlab(paste0("PC1: ", PR_pcavar.dbl[1], "% var. ")) +
  ylab(paste0("PC2: ", PR_pcavar.dbl[2], "% var. ")) +
  ggtheme.jj() +
  scale_color_manual(values = RNAi_cols.chr,
    breaks = c("LACZ", "GAF", "BAP170", "NURF301", "NURF301BAP170"),
    labels = c("LACZ", "GAF", "BAP", "NURF", "N+B")) +
  scale_x_continuous(limits = c(-50, 50),
    expand = c(0, 0),
    breaks = c(-50, -25, 0, 25, 50)) +
  scale_y_continuous(limits = c(-30, 30),
    expand = c(0, 0),
    breaks = c(-30, -15, 0, 15, 30))

```

```

# Panel C - E23 browser shot
E23.gr <- genes.gr[which(genes.gr$tx_name == "E23-RC")]

# moving end of gene in (to condense plot)
start(E23.gr) <- start(E23.gr) + 19000

# Plotting (see file browserPlotR.R for more detail)
F1C <- browser_plotter.jj(
  E23.gr,
  list("PROseq" = PROseq_normed_merged.lst,
        "ATACDHS" = ATACDHS_normed_merged.lst,
        "ChIPseq" = ChIPnew_merged.lst),
  labs = c("LACZ", "GAF", "BAP", "NURF", "N+B",
            "LACZ", "GAF", "BAP", "NURF", "N+B",
            "LACZ RNAi GAF ChIP", "GAF RNAi GAF ChIP",
            "BAP RNAi GAF ChIP"),
  scale_bar_size = 1000,
  pad_left = 0,
  pad_right = 500,
  binsize = 1,
  bin_FUN = mean,
  .expand_ranges = list(
    "PROseq" = FALSE,
    "ATACDHS" = TRUE,
    "ChIPseq" = TRUE)
)

# Panel D - MA plot of promoter region GAF vs LACZ
# Get df of counts for running DESeq2
PR.df <- data.frame(PR.tbl, row.names = "tx_name")
PR_col_data.df <- data.frame(
  row.names = colnames(PR.df),
  RNAi = c(
    "BAP170",
    "BAP170",
    "GAF",
    "GAF",
    "LACZ",
    "LACZ",
    "NURFBAP",
    "NURFBAP",
    "NURF301",
    "NURF301"
  )
)

# Getting DESeq results object, GAF vs LACZ
PR_GAF.res <-
  subset_DESeq.jj(
    PR.df,
    PR_col_data.df,
    scale_facts_PROseq.dbl,

```



```

c(3:6),
c("RNAi", "GAF", "LACZ"),
~RNAi)

# Converting results object to df
PR_GAFres.df <- as.data.frame(PR_GAF.res)
PR_GAFres.df$tx_name <- row.names(PR_GAFres.df)

# Writing results object as file
write_tsv(PR_GAFres.df, "DESeq_results/GAF_PR_res.tsv")

# Plotting
F1D <- maplot.jj(PR_GAF.res)

# Panel E - GAF-dependent promoters ATAC-seq (<120bp) metaprofile

# Getting list of GAF-dependent promoters (FDR 0.01)
gafDepPR.gr <-
  genes.gr[which(
    genes.gr$tx_name %in%
      (PR_GAFres.df %>%
        filter(padj < 0.01 & log2FoldChange < 0))$tx_name
  )]

# Getting counts matrix of ATAC data
gafDepPR_ATACDHSSub.mat <-
  metaSubsample(
    dataset.gr = ATACDHS_normed_merged.lst,
    regions.gr = promoters(gafDepPR.gr, 500, 500),
    binsize = 1,
    first.output.xval = -500,
    expand_ranges = TRUE
  )

# Factoring sample.name column
gafDepPR_ATACDHSSub.mat$sample.name <-
  factor(
    gafDepPR_ATACDHSSub.mat$sample.name,
    levels = names(ATACDHS_normed_merged.lst)
  )

# Plotting
F1E <- ggMetaplot(gafDepPR_ATACDHSSub.mat) +
  gg_color_scale_RNAi_ATACDHS +
  gg_fill_scale_RNAi_ATACDHS +
  scale_x_continuous(
    limits = c(-500, 500),
    expand = c(0,0)
  ) +
  scale_y_continuous(
    limits = c(0, 200),
    expand = c(0,0)
  )

```

```

    )+
    ggtitle("ATAC < 120 bp")

# Panel F - GAF-dependent promoters PRO-seq (<120bp) metaprofile
# Getting subsampled count matrices
gafDepPR_proSeqSub.mat <-
  metaSubsample(
    dataset.gr = PROseq_normed_merged.lst,
    regions.gr = promoters(gafDepPR.gr, 0, 150),
    binsize = 2,
    first.output.xval = 0,
    expand_ranges = FALSE
  )

gafDepGB_proSeqSub.mat <-
  metaSubsample(
    dataset.gr = PROseq_normed_merged.lst,
    regions.gr = promoters(gafDepPR.gr, 0, 1500),
    binsize = 20,
    first.output.xval = 0,
    expand_ranges = FALSE
  )

# Filling sample.name column and factoring
gafDepPR_proSeqSub.mat$sample.name <-
  factor(
    gafDepPR_proSeqSub.mat$sample.name,
    levels = names(PROseq_normed_merged.lst)
  )

gafDepGB_proSeqSub.mat$sample.name <-
  factor(
    gafDepGB_proSeqSub.mat$sample.name,
    levels = names(PROseq_normed_merged.lst)
  )

# Plotting promoter
F1FHa <- ggMetaplot(gafDepPR_proSeqSub.mat)+
  gg_color_scale_RNAi_PROseq +
  gg_fill_scale_RNAi_PROseq +
  scale_x_continuous(
    limits = c(0, 150),
    expand = c(0,0),
    breaks = c(0, 50, 100, 150)
  ) +
  scale_y_continuous(
    breaks = c(0, 50, 100)
  ) +
  coord_cartesian(
    ylim = c(0, 100),
    expand = 0,

```

```

    )+
    ggtitle("PRO-seq")

# Plotting gene body
F1FHb <- ggMetaplot(gafDepGB_proSeqSub.mat)+
  gg_color_scale_RNAi_PROseq +
  gg_fill_scale_RNAi_PROseq +
  scale_y_continuous(
    breaks = c(0, 0.25, 0.5, 0.75)
  )+
  scale_x_continuous(
    breaks = c(seq(150, 1500, length.out = 4))
  )+
  coord_cartesian(
    xlim = c(150, 1500),
    ylim = c(0, 0.75),
    expand = 0,
  ) +
  ylab(NULL)

# Panel G - GAF ChIP-seq metaplots after GAF/LACZ/BAP RNAi
gafDepPR_GAFChIP.mat <-
  metaSubsample(
    dataset.gr = ChIPnew_merged.lst,
    regions.gr = promoters(gafDepPR.gr, 500, 500),
    binsize = 10,
    first.output.xval = -500,
    expand_ranges = TRUE
  )

F1G <- ggMetaplot(gafDepPR_GAFChIP.mat)+
  gg_color_scale_RNAi_PROseq +
  gg_fill_scale_RNAi_PROseq +
  scale_x_continuous(
    limits = c(-500, 500),
    expand = c(0,0),
    breaks = c(-500, -250, 0, 250, 500)
  ) +
  scale_y_continuous(
    limits = c(0, 0.6),
    breaks = c(0, 0.3, 0.6),
    expand = c(0,0)
  )+
  ggtitle("GAF ChIP-seq")
ggsave("Fig1G.pdf", F1G, width = 3.5, height = 1.6)

# Panel H - Duplicate of E - easier to remove irrelevant lines from each later

# Panel I - GAF PR l2FC vs PBAP and NURF scatterplot
# Running DESeq2 for BAP and NURF vs LACZ control
PR_BAP.res <-
  subset_DESeq.jj(

```

```

PR.df,
PR_col_data.df,
scale_facts_PROseq.dbl,
c(1, 2, 5, 6),
c("RNAi", "BAP170", "LACZ"),
~ RNAi
)

PR_NURF.res <-
subset_DESeq.jj(
PR.df,
PR_col_data.df,
scale_facts_PROseq.dbl,
c(5, 6, 9, 10),
c("RNAi", "NURF301", "LACZ"),
~ RNAi
)

# Plotting GAF vs BAP
F1Ha <- fc_corr.jj(PR_GAF.res, PR_BAP.res)+
coord_cartesian(xlim = c(-8, 8), ylim = c(-8, 8), expand = FALSE)+
xlab("GAF PR 12FC")+
ylab("BAP170 PR 12FC")

# Plotting GAF vs NURF
F1Hb <- fc_corr.jj(PR_GAF.res, PR_NURF.res)+
coord_cartesian(xlim = c(-8, 8), ylim = c(-8, 8), expand = FALSE)+
xlab("GAF PR 12FC")+
ylab("BAP170 PR 12FC")

# Layout for plotting
F1_layout <- "
#CDK
AEF#
BGH#
BIJ#
"

# Combining all panels
F1_all <- F1B + F1C + F1D + F1E + F1FHa +
F1FHb + F1FHa + F1FHb + F1Ha + F1Hb + F1G +
plot_layout(guides = "collect",
design = F1_layout) +
plot_annotation(tag_levels = 'A') &
theme(plot.tag = element_text(family = "Helvetica", face = "bold", size = 12))

```

Figure S1

A - RNA-seq knockdown efficiency

B/C - Western blots showing knockdown efficiency of various factors. No analysis needed, assembled in Illustrator

```

# Panel A - Knockdown efficiency by RNA-seq

# Normalizing RNA-seq counts using ERCC counts
# scale_factor = minimum(mapped_ERCC) / sample(mapped_ERCC)
RNA_normed.tbl <- RNA.tbl
for (i in colnames(RNA.tbl[, -1])) {
  RNA_normed.tbl[, i] <- RNA.tbl[, i] / scale_facts_RNAseq.lst[[i]]
}

# Filtering just rows for the three factors that were knocked down
# and dividing each by their transcript abundance in the lacZ condition
# to get a fold change value
KD.tbl <-
  RNA_normed.tbl[
    which(
      RNA_normed.tbl$tx_name %in% c("Trl-RA", "Bap170-RA", "E(bx)-RC")
    ),
  ] %>%
  transmute(
    "BAP_Rep1_fc" = (BAP170_RNAseq_Rep1 / LACZ_RNAseq_Rep1),
    "BAP_Rep2_fc" = (BAP170_RNAseq_Rep2 / LACZ_RNAseq_Rep2),
    "GAF_Rep1_fc" = (GAF_RNAseq_Rep1 / LACZ_RNAseq_Rep1),
    "GAF_Rep2_fc" = (GAF_RNAseq_Rep2 / LACZ_RNAseq_Rep2),
    "NURFBAP_Rep1_fc" = (NURF301BAP170_RNAseq_Rep1 / LACZ_RNAseq_Rep1),
    "NURFBAP_Rep2_fc" = (NURF301BAP170_RNAseq_Rep2 / LACZ_RNAseq_Rep2),
    "NURF_Rep1_fc" = (NURF301_RNAseq_Rep1 / LACZ_RNAseq_Rep1),
    "NURF_Rep2_fc" = (NURF301_RNAseq_Rep2 / LACZ_RNAseq_Rep2),
    "tx_name" = tx_name
  ) %>%
  mutate("tx_name" =
    if_else(tx_name == "Trl-RA", "GAF",
      if_else(tx_name == "Bap170-RA", "BAP",
        if_else(tx_name == "E(bx)-RC", "NURF",
          "no_match_found")))) %>%
  gather("target", "FC", -tx_name) %>%
  separate(target, into = c("target", NA, NA))

# Only retaining rows were target matches RNAi
KD.tbl <-
  KD.tbl[which(mapply(grepl, KD.tbl$tx_name, KD.tbl$target)), ]

# Concatenating ID and target for plotting
KD.tbl$tx_name_target <-
  paste(KD.tbl$tx_name, KD.tbl$target, sep = "_")

# Summarizing by group and calculating SEM
KD.tbl <- group_by(KD.tbl, tx_name_target) %>%
  dplyr::summarise(mean_FC = mean(FC),
    sd_FC = sd(FC))

# Factoring levels for plotting
KD.tbl$tx_name_target <-

```

```

factor(
  KD.tbl$tx_name_target,
  levels = c(
    "GAF_GAF",
    "BAP_BAP",
    "NURF_NURF",
    "BAP_NURFBAP",
    "NURF_NURFBAP"
  )
)

# Plotting barplot
S1A <- ggplot(KD.tbl,
  aes(x = tx_name_target, y = mean_FC, fill = tx_name_target)
)+
  geom_col(show.legend = F)+
  geom_errorbar(
    aes(ymin = mean_FC - sd_FC, ymax = mean_FC + sd_FC), width = 0.1
  )+
  ggtheme.jj() +
  scale_fill_manual(values = c(
    "#66CCEE",
    "#228833", # Manual color scaling because
    "#CCBB44", # the NURF+BAP condition has
    "#AA3377", # two values in this ploy and must
    "#AA3377" # be entered twice
  )) +
  scale_y_continuous(limits = c(0, 1.0), expand = c(0, 0))+
  scale_x_discrete(
    labels = c("GAF", "BAP170", "NURF301", "N+B-BAP", "N+B-NURF")
  ) +
  ylab("Fraction mRNA") +
  xlab("RNAi")+
  theme(
    axis.line.x = element_blank(),
    axis.ticks.x = element_blank(),
    axis.text.x = element_text(angle = 45, hjust = 1)
  )

# Filtering out genes from the RNA-seq counts table that
# aren't in the PRO-seq table (blacklisted due to too much
# US transcription). This has to wait until now because
# Trl-RA is one of the genes that gets filtered
RNA.tbl <-
  RNA.tbl[which(!(RNA.tbl$tx_name %in% blacklist.chr)),]
RNA_normed.tbl <-
  RNA_normed.tbl[which(!(RNA_normed.tbl$tx_name %in% blacklist.chr)),]

```

Figure S2

A - PCA PRO-seq GB
 B - PCA RNA-seq genes

C - PCA ATAC-seq peaks
D - PCA ATAC-seq promoters

```
# Panel A - PCA of PRO-seq signal in gene body regions
# Formatting count matrix for DESeq2. Rownames = gene names
GB.df <- data.frame(GB.tbl, row.names = "tx_name")

# colData object for DESeq2
col_data.df <- data.frame(
  row.names = colnames(GB.df),
  RNAi = c(
    "BAP170",
    "BAP170",
    "GAF",
    "GAF",
    "LACZ",
    "LACZ",
    "NURF301BAP170",
    "NURF301BAP170",
    "NURF301",
    "NURF301"
  )
)

# Construction of DESeq2 object
GB.dds <- DESeqDataSetFromMatrix(countData = GB.df,
                                colData = col_data.df,
                                design = ~ RNAi)

# Adding spike-in scale factors
sizeFactors(GB.dds) <- scale_facts_PROseq.dbl

# Running DESeq2
GB.dds <- DESeq(GB.dds)

# Variance stabilizing log transform
GB.rld <- rlog(GB.dds, blind = F)

# Getting PCA data
GB.pca <-
  plotPCA(GB.rld,
    intgroup = c("RNAi"),
    returnData = TRUE)

# Getting % Var explained by each PC
GB_pcavar.dbl <- round(100 * attr(GB.pca, "percentVar"))

# Factoring for plotting
GB.pca$RNAi <-
  factor(GB.pca$RNAi,
    levels =
      c("LACZ", "GAF", "BAP170", "NURF301", "NURF301BAP170")
  )
```

```

# Plotting
S2A <- ggplot(GB.pca, aes(PC1, PC2, color = RNAi)) +
  geom_point(size = 2, show.legend = T) +
  xlab(paste0("PC1: ", GB_pcavar.dbl[1], "% var. ")) +
  ylab(paste0("PC2: ", GB_pcavar.dbl[2], "% var. ")) +
  ggtheme.jj() +
  scale_color_manual(
    values = RNAi_cols.chr,
    breaks = c("LACZ", "GAF", "BAP170", "NURF301", "NURF301BAP170"),
    labels = c("LACZ", "GAF", "BAP", "NURF", "N+B")
  ) +
  scale_x_continuous(
    limits = c(-20, 30),
    expand = c(0, 0),
    breaks = c(-20, -10, 0, 10, 20, 30)
  ) +
  scale_y_continuous(
    limits = c(-20, 30),
    expand = c(0, 0),
    breaks = c(-20, -10, 0, 10, 20, 30)
  ) +
  ggtitle("Gene Body PRO-seq")

# Panel B - PCA of RNA-seq counts per gene
# Formatting count matrix for DESeq2. Rownames = gene names
RNA.tbl <- RNA.tbl[which(!(RNA.tbl$tx_name %in% blacklist.chr)), ]
RNA.df <- data.frame(RNA.tbl, row.names = "tx_name")

# colData object for DESeq2
col_data.df <- data.frame(
  row.names = colnames(RNA.df),
  RNAi = c(
    "BAP170",
    "BAP170",
    "GAF",
    "GAF",
    "LACZ",
    "LACZ",
    "NURF301BAP170",
    "NURF301BAP170",
    "NURF301",
    "NURF301"
  )
)

# construction of DESeq2 object
RNA.dds <- DESeqDataSetFromMatrix(countData = RNA.df,
                                   colData = col_data.df,
                                   design = ~ RNAi)

# Adding spike-in scale factors
sizeFactors(RNA.dds) <- scale_factors_RNAseq.dbl

```



```

# Running DESeq2
RNA.dds <- DESeq(RNA.dds)

# Variance stabilizing log transform
RNA.rld <- rlog(RNA.dds, blind = F)

# Getting PCA data
RNA.pca <-
  plotPCA(RNA.rld,
    intgroup = c("RNAi"),
    returnData = TRUE)
# Getting % Var explained by each PC
RNA_pcavar.dbl <- round(100 * attr(RNA.pca, "percentVar"))

# Factoring for plotting
RNA.pca$RNAi <-
  factor(RNA.pca$RNAi,
    levels =
      c("LACZ", "GAF", "BAP170", "NURF301", "NURF301BAP170")
  )

# Plotting
S2B <- ggplot(RNA.pca, aes(PC1, PC2, color = RNAi)) +
  geom_point(size = 2) +
  xlab(paste0("PC1: ", RNA_pcavar.dbl[1], "% var. ")) +
  ylab(paste0("PC2: ", RNA_pcavar.dbl[2], "% var. ")) +
  ggtheme.jj() +
  scale_color_manual(
    values = RNAi_cols.chr,
    breaks = c("LACZ", "GAF", "BAP170", "NURF301", "NURF301BAP170"),
    labels = c("LACZ", "GAF", "BAP", "NURF", "N+B")
  ) +
  scale_x_continuous(
    limits = c(-20, 20),
    expand = c(0, 0),
    breaks = c(-20, -10, 0, 10, 20)
  ) +
  scale_y_continuous(
    limits = c(-15, 15),
    expand = c(0, 0),
    breaks = c(-15, 0, 15)
  ) +
  ggtitle("3'RNA-seq")

# Panel C - PCA of ATAC-seq counts in peaks called using macs2
# Formatting count matrix for DESeq2. Rownames = gene names
ATAC_peaks.df <- data.frame(ATAC_peaks.tbl, row.names = "peak_name")

# colData object for DESeq2
col_data.df <- data.frame(
  row.names = colnames(ATAC_peaks.df),
  RNAi = c(
    "BAP170",

```

```

    "BAP170",
    "GAF",
    "GAF",
    "LACZ",
    "LACZ",
    "NURF301BAP170",
    "NURF301BAP170",
    "NURF301",
    "NURF301"
  )
)

# Construction of DESeq2 object
ATAC_peaks.dds <- DESeqDataSetFromMatrix(countData = ATAC_peaks.df,
                                         colData = col_data.df,
                                         design = ~ RNAi)

# Using DESeq2 internal normalization for ATAC-seq data

# Running DESeq2
ATAC_peaks.dds <- DESeq(ATAC_peaks.dds)

# Variance stabilizing log transform
ATAC_peaks.rld <- rlog(ATAC_peaks.dds, blind = F)

# Getting PCA data
ATAC_peaks.pca <-
  plotPCA(ATAC_peaks.rld,
          intgroup = c("RNAi"),
          returnData = TRUE)
# Getting % Var explained by each PC
ATAC_peaks_pcavar.dbl <-
  round(100 * attr(ATAC_peaks.pca, "percentVar"))

# Factoring for plotting
ATAC_peaks.pca$RNAi <-
  factor(ATAC_peaks.pca$RNAi,
        levels =
          c("LACZ", "GAF", "BAP170", "NURF301", "NURF301BAP170")
  )

# Plotting
S2C <- ggplot(ATAC_peaks.pca, aes(PC1, PC2, color = RNAi)) +
  geom_point(size = 2, show.legend = T) +
  xlab(paste0("PC1: ", ATAC_peaks_pcavar.dbl[1], "% var. ")) +
  ylab(paste0("PC2: ", ATAC_peaks_pcavar.dbl[2], "% var. ")) +
  ggtheme.jj() +
  scale_color_manual(
    values = RNAi_cols.chr,
    breaks = c("LACZ", "GAF", "BAP170", "NURF301", "NURF301BAP170"),
    labels = c("LACZ", "GAF", "BAP", "NURF", "N+B")
  ) +
  scale_x_continuous(

```

```

limits = c(-40, 40),
expand = c(0, 0),
breaks = c(-40,-20, 0, 20, 40)
) +
scale_y_continuous(
  limits = c(-20, 20),
  expand = c(0, 0),
  breaks = c(-20,-10, 0, 10, 20)
) +
ggtitle("ATAC-seq (peaks)")

# Panel D - PCA of ATAC-seq counts in promoter regions
# Getting counts matrix
ATAC_PR.df <- data.frame(ATAC_PR.tbl, row.names = "tx_name")

# colData object for DESeq2
col_data.df <- data.frame(
  row.names = colnames(ATAC_PR.df),
  RNAi = c(
    "BAP170",
    "BAP170",
    "GAF",
    "GAF",
    "LACZ",
    "LACZ",
    "NURF301BAP170",
    "NURF301BAP170",
    "NURF301",
    "NURF301"
  )
)

# construction of DESeq2 object
ATAC_PR.dds <- DESeqDataSetFromMatrix(countData = ATAC_PR.df,
                                      colData = col_data.df,
                                      design = ~ RNAi)

# Using DESeq2 internal normalization for ATAC-seq data

# Running DESeq2
ATAC_PR.dds <- DESeq(ATAC_PR.dds)

# Variance stabilizing log transform
ATAC_PR.rld <- rlog(ATAC_PR.dds, blind = F)

# Getting PCA data
ATAC_PR.pca <-
  plotPCA(ATAC_PR.rld,
          intgroup = c("RNAi"),
          returnData = TRUE)

# Getting % Var explained by each PC
ATAC_PR_pcavar.dbl <- round(100 * attr(ATAC_PR.pca, "percentVar"))

```

```

# Factoring for plotting
ATAC_PR.pca$RNAi <-
  factor(ATAC_PR.pca$RNAi,
    levels = c("LACZ", "GAF", "BAP170", "NURF301", "NURF301BAP170"))

# Plotting
S2D <- ggplot(ATAC_PR.pca, aes(PC1, PC2, color = RNAi)) +
  geom_point(size = 2) +
  xlab(paste0("PC1: ", ATAC_PR_pcavar.dbl[1], "% var. ")) +
  ylab(paste0("PC2: ", ATAC_PR_pcavar.dbl[2], "% var. ")) +
  ggtheme.jj() +
  scale_color_manual(values = RNAi_cols.chr,
    breaks = c("LACZ", "GAF", "BAP170", "NURF301", "NURF301BAP170"),
    labels = c("LACZ", "GAF", "BAP", "NURF", "N+B")) +
  scale_x_continuous(limits = c(-25, 25),
    expand = c(0,0),
    breaks = c(-25, -12.5, 0, 12.5, 25)) +
  scale_y_continuous(limits = c(-15, 15),
    expand = c(0,0),
    breaks = c(-15, -7.5, 0, 7.5, 15)) +
  ggtitle("ATAC-seq (promoters)")

# Combining panels
S2_all <- (S2A + S2B) / (S2C + S2D) +
  plot_layout(guides = 'collect') +
  plot_annotation(tag_levels = 'A') &
  theme(plot.tag = element_text(family = "Helvetica", face = "bold", size = 12))

# Saving ATACseq scale factors for later
scale_facts_ATACDHS.dbl <- sizeFactors(ATAC_PR.dds)
scale_facts_ATACDHS.lst <- setNames(as.list(scale_facts_ATACDHS.dbl), samples_ATACDHS.chr)

```

Figure S3

- A - scatter plots PRO-seq GB raw
- B - scatter plots PRO-seq GB normalized
- C - scatter plots PRO-seq PR raw
- D - scatter plots PRO-seq PR normalized

```

# Panel A - scatter plots PRO-seq GB raw
S3A <- rep_corr_scatter.jj(
  df = GB.tbl,
  sep_colnames_into = c("condition", NA, "rep")) +
  scale_y_log10(limits = c(1, 1000000), expand = c(0,0),
    breaks = c(1, 10, 100, 1000, 10000, 100000, 1000000),
    labels = c(0, 1, 2, 3, 4, 5, 6)) +
  scale_x_log10(limits = c(1, 1000000), expand = c(0,0),
    breaks = c(1, 10, 100, 1000, 10000, 100000, 1000000),
    labels = c(0, 1, 2, 3, 4, 5, 6)) +
  facet_grid(. ~ condition) +
  xlab("log10 Rep. 1") +
  ylab("log10 Rep. 2") +

```

```

ggtitle("PRO-seq Gene Body Raw Counts")+
guides(fill = guide_colourbar(barwidth = 0.5, barheight = 5))

# Panel B - scatter plots PRO-seq GB normalized
S3B <- rep_corr_scatter.jj(
  df = GB_normed.tbl,
  sep_colnames_into = c("condition", NA, "rep")) +
  scale_y_log10(limits = c(1, 1000000), expand = c(0,0),
               breaks = c(1, 10, 100, 1000, 10000, 100000, 1000000),
               labels = c(0, 1, 2, 3, 4, 5, 6))+
  scale_x_log10(limits = c(1, 1000000), expand = c(0,0),
               breaks = c(1, 10, 100, 1000, 10000, 100000, 1000000),
               labels = c(0, 1, 2, 3, 4, 5, 6))+
  facet_grid(.~ condition)+
  xlab("log10 Rep. 1") +
  ylab("log10 Rep. 2") +
  ggtitle("PRO-seq Gene Body Spike-In Normalized Counts")+
  guides(fill = guide_colourbar(barwidth = 0.5, barheight = 5))

# Panel C - scatter plots PRO-seq PR raw
S3C <- rep_corr_scatter.jj(
  df = PR.tbl,
  sep_colnames_into = c("condition", NA, "rep")) +
  scale_y_log10(limits = c(1, 100000), expand = c(0,0),
               breaks = c(1, 10, 100, 1000, 10000, 100000),
               labels = c(0, 1, 2, 3, 4, 5))+
  scale_x_log10(limits = c(1, 100000), expand = c(0,0),
               breaks = c(1, 10, 100, 1000, 10000, 100000),
               labels = c(0, 1, 2, 3, 4, 5))+
  facet_grid(.~ condition)+
  xlab("log10 Rep. 1") +
  ylab("log10 Rep. 2") +
  ggtitle("PRO-seq Promoter Raw Counts")+
  guides(fill = guide_colourbar(barwidth = 0.5, barheight = 5))

# Panel D - scatter plots PRO-seq PR normalized
S3D <- rep_corr_scatter.jj(
  df = PR_normed.tbl,
  sep_colnames_into = c("condition", NA, "rep")) +
  scale_y_log10(limits = c(1, 100000), expand = c(0,0),
               breaks = c(1, 10, 100, 1000, 10000, 100000),
               labels = c(0, 1, 2, 3, 4, 5))+
  scale_x_log10(limits = c(1, 100000), expand = c(0,0),
               breaks = c(1, 10, 100, 1000, 10000, 100000),
               labels = c(0, 1, 2, 3, 4, 5))+
  facet_grid(.~ condition)+
  xlab("log10 Rep. 1") +
  ylab("log10 Rep. 2") +
  ggtitle("PRO-seq Promoter Spike-In Normalized Counts")+
  guides(fill = guide_colourbar(barwidth = 0.5, barheight = 5))

# Combining panels
S3_all <- S3A / S3B / S3C / S3D +

```

```

plot_annotation(tag_levels = 'A') &
theme(plot.tag = element_text(family = "Helvetica",
                              face = "bold",
                              size = 12))

```

Figure S4

- A - scatter plots RNA-seq raw
- B - scatter plots RNA-seq normalized
- C - scatter plots ATAC-seq raw
- D - scatter plots ATAC-seq normalized

```

# Panel A - scatter plots RNA-seq raw
S4A <- rep_corr_scatter.jj(
  df = RNA.tbl,
  sep_colnames_into = c("condition", NA, "rep")) +
  scale_y_log10(limits = c(1, 100000), expand = c(0,0),
               breaks = c(1, 10, 100, 1000, 10000, 100000),
               labels = c(0, 1, 2, 3, 4, 5))+
  scale_x_log10(limits = c(1, 100000), expand = c(0,0),
               breaks = c(1, 10, 100, 1000, 10000, 100000),
               labels = c(0, 1, 2, 3, 4, 5))+
  facet_grid(~ condition)+
  xlab("log10 Rep. 1") +
  ylab("log10 Rep. 2") +
  ggtitle("3'RNA-seq Raw Counts")+
  guides(fill = guide_colourbar(barwidth = 0.5, barheight = 5))

# Panel B - scatter plots RNA-seq normalized
S4B <- rep_corr_scatter.jj(
  df = RNA_normed.tbl,
  sep_colnames_into = c("condition", NA, "rep")) +
  scale_y_log10(limits = c(1, 100000), expand = c(0,0),
               breaks = c(1, 10, 100, 1000, 10000, 100000),
               labels = c(0, 1, 2, 3, 4, 5))+
  scale_x_log10(limits = c(1, 100000), expand = c(0,0),
               breaks = c(1, 10, 100, 1000, 10000, 100000),
               labels = c(0, 1, 2, 3, 4, 5))+
  facet_grid(~ condition)+
  xlab("log10 Rep. 1") +
  ylab("log10 Rep. 2") +
  ggtitle("3'RNA-seq Spike-In Normalized Counts")+
  guides(fill = guide_colourbar(barwidth = 0.5, barheight = 5))

# Panel C - scatter plots ATAC-seq raw
S4C <- rep_corr_scatter.jj(
  df = ATAC_PR.tbl,
  sep_colnames_into = c("condition", NA, "rep")) +
  scale_y_log10(limits = c(1, 1000000), expand = c(0,0),
               breaks = c(1, 10, 100, 1000, 10000, 100000, 1000000),
               labels = c(0, 1, 2, 3, 4, 5, 6))+
  scale_x_log10(limits = c(1, 1000000), expand = c(0,0),
               breaks = c(1, 10, 100, 1000, 10000, 100000, 1000000),

```

```

        labels = c(0, 1, 2, 3, 4, 5, 6))+
facet_grid(~ condition)+
xlab("log10 Rep. 1") +
ylab("log10 Rep. 2") +
ggtitle("ATAC-seq Promoter Raw Counts")+
guides(fill = guide_colourbar(barwidth = 0.5, barheight = 5))

# Panel D - scatter plots ATAC-seq normalized

## Normalizing ATAC-seq data
ATAC_PR_normed.tbl <- ATAC_PR.tbl
for (i in colnames(ATAC_PR.tbl[, -1])) {
  ATAC_PR_normed.tbl[, i] <- ATAC_PR.tbl[, i] / scale_facts_ATACDHS.lst[[i]]
}

# Plotting
S4D <- rep_corr_scatter.jj(
  df = ATAC_PR_normed.tbl,
  sep_colnames_into = c("condition", NA, "rep")) +
  scale_y_log10(limits = c(1, 1000000), expand = c(0,0),
               breaks = c(1, 10, 100, 1000, 10000, 100000, 1000000),
               labels = c(0, 1, 2, 3, 4, 5, 6))+
  scale_x_log10(limits = c(1, 1000000), expand = c(0,0),
               breaks = c(1, 10, 100, 1000, 10000, 100000, 1000000),
               labels = c(0, 1, 2, 3, 4, 5, 6))+
  facet_grid(~ condition)+
  xlab("log10 Rep. 1") +
  ylab("log10 Rep. 2") +
  ggtitle("ATAC-seq Promoter Normalized Counts")+
  guides(fill = guide_colourbar(barwidth = 0.5, barheight = 5))

# Combining panels
S4_all <- S4A / S4B / S4C / S4D + plot_annotation(tag_levels = 'A') &
  theme(plot.tag = element_text(family = "Helvetica",
                                face = "bold", size = 12))

```

Figure 2

A - DREME logos from GAF bound unchanged PR vs GAF bound changed PR
 B - ChIP-seq, PRO-seq and ATAC-seq at classified GAF promoters

```

# A - DREME logos from GAF bound unchanged PR vs GAF bound changed PR
# Defining set of GAF bound genes, classified by whether they
# have decreased pausing upon GAF knockdown

# Getting list of genes and distance of each promoter
# to nearest GAF peak
gafBoundGenes.gr <- genes.gr
gafBoundGenes.gr$distToGAFpeak <-
  mcols(distanceToNearest(
    promoters(gafBoundGenes.gr, 500, 0),
    GAF_peaks.gr))$distance

```

```

# subsetting to genes where dist == 0, so GAF peak within TSS-500 to TSS)
gafBoundGenes.gr <-
  gafBoundGenes.gr[gafBoundGenes.gr$distToGAFpeak == 0, ]

# Dividing based on behavior of promoter upon GAF-RNAi
gafBoundGenes_gafDepPR.gr <-
  gafBoundGenes.gr[gafBoundGenes.gr$tx_name %in% gafDepPR.gr$tx_name, ]
gafBoundGenes_gafIndPR.gr <-
  gafBoundGenes.gr[!gafBoundGenes.gr$tx_name %in% gafDepPR.gr$tx_name, ]

# Use these as input for DREME (external) and insert motifs into figure
write_tsv(
  as.data.frame(
    promoters(
      gafBoundGenes_gafDepPR.gr, 500, 0)),
  "bed/gafBoundGenes_gafDepPR.bed")
write_tsv(
  as.data.frame(
    promoters(
      gafBoundGenes_gafIndPR.gr, 500, 0)),
  "bed/gafBoundGenes_gafIndPR.bed")

# Getting GAF ChIP signal intensity for
# gafBoundGenes_gafDepPR.gr (for sorting later)
gafBoundGenes_gafDepPR.gr$GAF_signal <-
  getCountsByRegions(
    ChIP.lst$GAF_ChIPseq,
    promoters(gafBoundGenes_gafDepPR.gr, 500, 0),
    expand_ranges = TRUE
  )

# B - ChIP-seq, PRO-seq and ATAC-seq at classified GAF promoters
# Adding total ChIP-seq signal for M1BP and BEAF32 to genes
gafBoundGenes_gafIndPR.gr$M1BP_signal <-
  getCountsByRegions(
    ChIP.lst$M1BP_ChIPseq,
    promoters(gafBoundGenes_gafIndPR.gr, 500, 0),
    expand_ranges = TRUE
  )

gafBoundGenes_gafIndPR.gr$BEAF32_signal <-
  getCountsByRegions(
    ChIP.lst$BEAF32_ChIPseq,
    promoters(gafBoundGenes_gafIndPR.gr, 500, 0),
    expand_ranges = TRUE
  )

# Defining classes, considering a promoter "bound" by each factor
# if it's in the top 25% of promoters for signal of that factor by ChIP-seq
gafBoundGenes_gafIndPR.gr$class <- ifelse(
  gafBoundGenes_gafIndPR.gr$M1BP_signal > quantile(
    gafBoundGenes_gafIndPR.gr$M1BP_signal)[[4]],
  ifelse(

```



```

gafBoundGenes_gafIndPR.gr$BEAF32_signal > quantile(
  gafBoundGenes_gafIndPR.gr$BEAF32_signal)[[4]],
  "Both",
  "M1BP"
),
}else(
  gafBoundGenes_gafIndPR.gr$BEAF32_signal > quantile(
    gafBoundGenes_gafIndPR.gr$BEAF32_signal)[[4]],
    "BEAF32",
    NA
  ))

# Dropping promoters that aren't either M1BP or BEAF32 bound
gafBoundGenes_gafIndPR.gr <-
  gafBoundGenes_gafIndPR.gr[!is.na(gafBoundGenes_gafIndPR.gr$class),]

# Concatenating and sorting GRs:
# 1 - GAFdepGAFbound sorted by GAF intensity
# 2 - GAFindGAFbound M1BP/BEAF32 bound sorted by M1BP intensity
# 3 - GAFindGAFbound M1BP bound sorted by M1BP intensity
# 4 - GAFindGAFbound BEAF32 bound sorted by BEAF32 intensity
gafDepPR_sorted.gr <-
  c(
    sort(gafBoundGenes_gafDepPR.gr,
      by = ~ GAF_signal, decreasing = TRUE),
    sort(
      gafBoundGenes_gafIndPR.gr[
        gafBoundGenes_gafIndPR.gr$class == "Both",
      ],
      by = ~ M1BP_signal, decreasing = TRUE),
    sort(
      gafBoundGenes_gafIndPR.gr[
        gafBoundGenes_gafIndPR.gr$class == "M1BP",
      ],
      by = ~ M1BP_signal, decreasing = TRUE),
    sort(
      gafBoundGenes_gafIndPR.gr[
        gafBoundGenes_gafIndPR.gr$class == "BEAF32",
      ],
      by = ~ BEAF32_signal, decreasing = TRUE)
  )

# Getting ChIP signal for hm
gafBoundGenesSorted_ChIP.mat <- cbind(
  getCountsByPositions(
    dataset.gr = ChIP.lst$GAF_ChIPseq,
    regions.gr = promoters(gafDepPR_sorted.gr, 500, 500),
    binsize = 10,
    expand_ranges = TRUE
  ),
  getCountsByPositions(
    dataset.gr = ChIP.lst$M1BP_ChIPseq,
    regions.gr = promoters(gafDepPR_sorted.gr, 500, 500),

```

```

    binsize = 10,
    expand_ranges = TRUE
  ),
  getCountsByPositions(
    dataset.gr = ChIP.lst$BEAF32_ChIPseq,
    regions.gr = promoters(gafDepPR_sorted.gr, 500, 500),
    binsize = 10,
    expand_ranges = TRUE
  )
)

# Setting up vectors to use for splitting heatmap rows and cols
row_split <-
  c(
    rep(
      "GAF",
      length(gafBoundGenes_gafDepPR.gr)
    ),
    rep(
      "Both",
      length(gafBoundGenes_gafIndPR.gr[
        gafBoundGenes_gafIndPR.gr$class == "Both",
        ])
    ),
    rep(
      "M1BP",
      length(gafBoundGenes_gafIndPR.gr[
        gafBoundGenes_gafIndPR.gr$class == "M1BP",
        ])
    ),
    rep(
      "BEAF32",
      length(gafBoundGenes_gafIndPR.gr[
        gafBoundGenes_gafIndPR.gr$class == "BEAF32",
        ])
    )
  )

row_split <- factor(row_split,
                    levels = c("GAF", "Both", "M1BP", "BEAF32"))
col_fun = colorRamp2(c(0, 50, 100), viridis(3))

# Plotting Heatmap
F2Ba <- Heatmap(
  gafBoundGenesSorted_ChIP.mat,
  cluster_columns = F,
  cluster_rows = F,
  col = col_fun,
  row_split = row_split
)

# Adding PRO-seq and ATAC-seq data to GR
gafDepPR_sorted.gr$LACZ_PROseq <-

```

```

getCountsByRegions(
  PROseq_normed_merged.lst$LACZ_PROseq,
  promoters(gafDepPR_sorted.gr,50,100),
  expand_ranges = FALSE
)

gafDepPR_sorted.gr$GAF_PROseq <-
  getCountsByRegions(
    PROseq_normed_merged.lst$GAF_PROseq,
    promoters(gafDepPR_sorted.gr,50,100),
    expand_ranges = FALSE
  )

gafDepPR_sorted.gr$BAP170_PROseq <-
  getCountsByRegions(
    PROseq_normed_merged.lst$BAP170_PROseq,
    promoters(gafDepPR_sorted.gr,50,100),
    expand_ranges = FALSE
  )

gafDepPR_sorted.gr$NURF301_PROseq <-
  getCountsByRegions(
    PROseq_normed_merged.lst$NURF301_PROseq,
    promoters(gafDepPR_sorted.gr,50,100),
    expand_ranges = FALSE
  )

gafDepPR_sorted.gr$NURF301BAP170_PROseq <-
  getCountsByRegions(
    PROseq_normed_merged.lst$NURF301BAP170_PROseq,
    promoters(gafDepPR_sorted.gr,50,100),
    expand_ranges = FALSE
  )

gafDepPR_sorted.gr$M1BPLACZ_PROseq <-
  getCountsByRegions(
    PROseq_M1BP.lst$LACZ,
    promoters(gafDepPR_sorted.gr,50,100),
    expand_ranges = FALSE
  )

gafDepPR_sorted.gr$M1BP_PROseq <-
  getCountsByRegions(
    PROseq_M1BP.lst$M1BP,
    promoters(gafDepPR_sorted.gr,50,100),
    expand_ranges = FALSE
  )

# Getting matrix of l2FC of each RNAi vs LACZ
# (adding pseudocount to prevent Inf/NA values)
gafDepPR_sorted_PROseql2FC.mat <- cbind(
  log2(
    (gafDepPR_sorted.gr$GAF_PROseq + 1) /

```

```

      (gafDepPR_sorted.gr$LACZ_PROseq + 1)
    ),
    log2(
      (gafDepPR_sorted.gr$BAP170_PROseq + 1) /
      (gafDepPR_sorted.gr$LACZ_PROseq + 1)
    ),
    log2(
      (gafDepPR_sorted.gr$NURF301_PROseq + 1) /
      (gafDepPR_sorted.gr$LACZ_PROseq + 1)
    ),
    log2(
      (gafDepPR_sorted.gr$NURF301BAP170_PROseq + 1) /
      (gafDepPR_sorted.gr$LACZ_PROseq + 1)
    )
  )
)

# Plotting Heatmap
col_fun_fc = colorRamp2(c(-2, 0, 2), c("#3B4992", "ghostwhite", "#BB0021"))
F2Bb <- Heatmap(
  gafDepPR_sorted_PROseq12FC.mat,
  cluster_columns = F,
  cluster_rows = F,
  col = col_fun_fc,
  row_split = row_split
)

# Plotting heatmap of M1BP l2fC pro-seq promoter
gafDepPR_sorted_M1BPPROseq12FC.mat <-
  log2((
    gafDepPR_sorted.gr$M1BP_PROseq + 1) /
    (gafDepPR_sorted.gr$M1BPLACZ_PROseq + 1)
  )

F2Bc <- Heatmap(
  gafDepPR_sorted_M1BPPROseq12FC.mat,
  cluster_columns = F,
  cluster_rows = F,
  col = col_fun_fc,
  row_split = row_split,
  column_labels = c("")
)

# Adding ATAC-seq data
gafDepPR_sorted.gr$LACZ_ATACseq <-
  getCountsByRegions(
    ATACDHS_normed_merged.lst$LACZ_ATACDHS,
    promoters(gafDepPR_sorted.gr, 250, 0),
    expand_ranges = TRUE
  )

gafDepPR_sorted.gr$GAF_ATACseq <-
  getCountsByRegions(

```

```

    ATACDHS_normed_merged.lst$GAF_ATACDHS,
    promoters(gafDepPR_sorted.gr, 250, 0),
    expand_ranges = TRUE
)

gafDepPR_sorted.gr$BAP170_ATACseq <-
  getCountsByRegions(
    ATACDHS_normed_merged.lst$BAP170_ATACDHS,
    promoters(gafDepPR_sorted.gr, 250, 0),
    expand_ranges = TRUE
  )

gafDepPR_sorted.gr$NURF301_ATACseq <-
  getCountsByRegions(
    ATACDHS_normed_merged.lst$NURF301_ATACDHS,
    promoters(gafDepPR_sorted.gr, 250, 0),
    expand_ranges = TRUE
  )

gafDepPR_sorted.gr$NURF301BAP170_ATACseq <-
  getCountsByRegions(
    ATACDHS_normed_merged.lst$NURF301BAP170_ATACDHS,
    promoters(gafDepPR_sorted.gr, 250, 0),
    expand_ranges = TRUE
  )

# Getting matrix of l2FC of each RNAi vs LACZ
gafDepPR_sorted_ATACseql2FC.mat <- cbind(
  log2(
    (gafDepPR_sorted.gr$GAF_ATACseq + 1) /
    (gafDepPR_sorted.gr$LACZ_ATACseq + 1)
  ),
  log2(
    (gafDepPR_sorted.gr$BAP170_ATACseq + 1) /
    (gafDepPR_sorted.gr$LACZ_ATACseq + 1)
  ),
  log2(
    (gafDepPR_sorted.gr$NURF301_ATACseq + 1) /
    (gafDepPR_sorted.gr$LACZ_ATACseq + 1)
  ),
  log2(
    (gafDepPR_sorted.gr$NURF301BAP170_ATACseq + 1) /
    (gafDepPR_sorted.gr$LACZ_ATACseq + 1)
  )
)

# Plotting Heatmap
F2Bd <- Heatmap(
  gafDepPR_sorted_ATACseql2FC.mat,
  cluster_columns = F,
  cluster_rows = F,
  col = col_fun_fc,
  row_split = row_split

```

)

Figure S5

A-D - MA plots promoter PRO-seq

E-H - MA plots GB PRO-seq

I-L - MA plots of RNA-seq

```
# A-D - MA plots promoter PRO-seq
# Running DESeq2 for remaining Promoter comparisons
PR_NURFBAP.res <-
  subset_DESeq.jj(
    PR.df, PR_col_data.df,
    scale_facts_PROseq.dbl,
    c(5,6,7,8),
    c("RNAi", "NURFBAP", "LACZ"),
    ~RNAi)

# Plotting
S5A <- maplot.jj(PR_GAF.res) # Moved to F1D, remove here
S5B <- maplot.jj(PR_BAP.res)
S5C <- maplot.jj(PR_NURF.res)
S5D <- maplot.jj(PR_NURFBAP.res)

# E-H - MA plots GB PRO-seq
# Running DESeq2 for all GB regions
GB.df <- data.frame(GB.tbl, row.names = "tx_name")
GB_col_data.df <- data.frame(
  row.names = colnames(GB.df),
  RNAi = c(
    "BAP170",
    "BAP170",
    "GAF",
    "GAF",
    "LACZ",
    "LACZ",
    "NURFBAP",
    "NURFBAP",
    "NURF301",
    "NURF301"
  )
)
GB_GAF.res <-
  subset_DESeq.jj(
    GB.df,
    GB_col_data.df,
    scale_facts_PROseq.dbl,
    c(3,4,5,6),
    c("RNAi", "GAF", "LACZ"),
    ~RNAi)
GB_BAP.res <-
  subset_DESeq.jj(
```

```

GB.df,
GB_col_data.df,
scale_facts_PROseq.dbl,
c(1,2,5,6),
c("RNAi", "BAP170", "LACZ"),
~RNAi)
GB_NURF.res <-
subset_DESeq.jj(
  GB.df,
  GB_col_data.df,
  scale_facts_PROseq.dbl,
  c(5,6,9,10),
  c("RNAi", "NURF301", "LACZ"),
  ~RNAi)
GB_NURFBAP.res <-
subset_DESeq.jj(
  GB.df,
  GB_col_data.df,
  scale_facts_PROseq.dbl,
  c(5,6,7,8),
  c("RNAi", "NURFBAP", "LACZ"),
  ~RNAi)

S5E <- maplot.jj(GB_GAF.res)
S5F <- maplot.jj(GB_BAP.res)
S5G <- maplot.jj(GB_NURF.res)
S5H <- maplot.jj(GB_NURFBAP.res)

# I-L - MA plots of RNA-seq
# Running DESeq2 for all RNA-seq
RNA.df <- data.frame(RNA.tbl, row.names = "tx_name")
RNA_col_data.df <- data.frame(
  row.names = colnames(RNA.df),
  RNAi = c(
    "BAP170",
    "BAP170",
    "GAF",
    "GAF",
    "LACZ",
    "LACZ",
    "NURFBAP",
    "NURFBAP",
    "NURF301",
    "NURF301"
  )
)
RNA_GAF.res <-
subset_DESeq.jj(
  RNA.df,
  RNA_col_data.df,
  scale_facts_RNAseq.dbl,
  c(3,4,5,6),
  c("RNAi", "GAF", "LACZ"),

```

```

~RNAi)
RNA_BAP.res <-
  subset_DESeq.jj(
    RNA.df,
    RNA_col_data.df,
    scale_facts_RNAseq.dbl,
    c(1,2,5,6),
    c("RNAi", "BAP170", "LACZ"),
    ~RNAi)
RNA_NURF.res <-
  subset_DESeq.jj(
    RNA.df,
    RNA_col_data.df,
    scale_facts_RNAseq.dbl,
    c(5,6,9,10),
    c("RNAi", "NURF301", "LACZ"),
    ~RNAi)
RNA_NURFBAP.res <-
  subset_DESeq.jj(
    RNA.df,
    RNA_col_data.df,
    scale_facts_RNAseq.dbl,
    c(5,6,7,8),
    c("RNAi", "NURFBAP", "LACZ"),
    ~RNAi)

S5I <- maplot.jj(RNA_GAF.res)
S5J <- maplot.jj(RNA_BAP.res)
S5K <- maplot.jj(RNA_NURF.res)
S5L <- maplot.jj(RNA_NURFBAP.res)

# Combining all panels
S5_all <- (S5A | S5B | S5C | S5D) /
  (S5E | S5F | S5G | S5H) /
  (S5I | S5J | S5K | S5L) +
  plot_annotation(tag_levels = 'A') &
  theme(plot.tag = element_text(
    family = "Helvetica",
    face = "bold",
    size = 12))

# Writing DESeq2 results to files
# Helper function
res_write_tsv.jj <-
  function(res, filename) {
    res.df <- as.data.frame(res)
    res.df$tx_name <- row.names(res.df)
    write_tsv(res.df, filename)
  }

res_write_tsv.jj(PR_GAF.res, "DESeq_results/GAF_PR_res.tsv")
res_write_tsv.jj(PR_BAP.res, "DESeq_results/BAP_PR_res.tsv")
res_write_tsv.jj(PR_NURF.res, "DESeq_results/NURF_PR_res.tsv")

```



```

res_write_tsv.jj(PR_NURFBAP.res, "DESeq_results/NURFBAP_PR_res.tsv")
res_write_tsv.jj(GB_GAF.res, "DESeq_results/GAF_GB_res.tsv")
res_write_tsv.jj(GB_BAP.res, "DESeq_results/BAP_GB_res.tsv")
res_write_tsv.jj(GB_NURF.res, "DESeq_results/NURF_GB_res.tsv")
res_write_tsv.jj(GB_NURFBAP.res, "DESeq_results/NURFBAP_GB_res.tsv")
res_write_tsv.jj(RNA_GAF.res, "DESeq_results/GAF_RNA_res.tsv")
res_write_tsv.jj(RNA_BAP.res, "DESeq_results/BAP_RNA_res.tsv")
res_write_tsv.jj(RNA_NURF.res, "DESeq_results/NURF_RNA_res.tsv")
res_write_tsv.jj(RNA_NURFBAP.res, "DESeq_results/NURFBAP_RNA_res.tsv")

```

Figure S6

A - Browser shot of glob1-RB

B - Browser shot of Cyp9c1

C - Browser shot of Fatp3-RA

D - Browser shot of geko RB

E - Browser shot of out-RA

Browser shots all made with R package found here:

https://github.com/JAJ256/browser_plot.R

Analysis in this paper performed with commit: 1352d5c

```

# A - glob1-RB
glob1.gr <- genes.gr[which(genes.gr$tx_name == "glob1-RB")]

# Moving end of gene in (to condense plot)
start(glob1.gr) <- start(glob1.gr) + 2000

# Plotting (see file browserPlotR.R for more detail)
S6A <- browser_plotter.jj(
  glob1.gr,
  list("PROseq" = PROseq_normed_merged.lst,
        "ATACDHS" = ATACDHS_normed_merged.lst,
        "ChIPseq" = ChIPnew_merged.lst),
  labs = c("LACZ", "GAF", "BAP", "NURF", "N+B",
            "LACZ", "GAF", "BAP", "NURF", "N+B",
            "LACZGAFChIP", "GAFGAFChIP", "BAPGAFChIP"),
  scale_bar_size = 1000,
  pad_left = 0,
  pad_right = 500,
  binsize = 1,
  bin_FUN = mean,
  .expand_ranges = list(
    "PROseq" = FALSE,
    "ATACDHS" = TRUE,
    "ChIPseq" = TRUE)
)

# B - Cyp9c1
Cyp9c1.gr <- genes.gr[which(genes.gr$tx_name == "Cyp9c1-RA")]

# Plotting (see file browserPlotR.R for more detail)
S6B <- browser_plotter.jj(

```

```

Cyp9c1.gr,
list("PROseq" = PROseq_normed_merged.lst,
      "ATACDHS" = ATACDHS_normed_merged.lst,
      "ChIPseq" = ChIPnew_merged.lst),
labs = c("LACZ", "GAF", "BAP", "NURF", "N+B",
          "LACZ", "GAF", "BAP", "NURF", "N+B",
          "LACZGAFChIP", "GAFGAFChIP", "BAPGAFChIP"),
scale_bar_size = 1000,
pad_left = 500,
pad_right = 0,
binsize = 1,
bin_FUN = mean,
.expand_ranges = list(
  "PROseq" = FALSE,
  "ATACDHS" = TRUE,
  "ChIPseq" = TRUE)
)

# C - Fatp3-RA
Fatp3.gr <- genes.gr[which(genes.gr$tx_name == "Fatp3-RA")]

# Plotting (see file browserPlotR.R for more detail)
S6C <- browser_plotter.jj(
  Fatp3.gr,
  list("PROseq" = PROseq_normed_merged.lst,
        "ATACDHS" = ATACDHS_normed_merged.lst,
        "ChIPseq" = ChIPnew_merged.lst),
  labs = c("LACZ", "GAF", "BAP", "NURF", "N+B",
            "LACZ", "GAF", "BAP", "NURF", "N+B",
            "LACZGAFChIP", "GAFGAFChIP", "BAPGAFChIP"),
  scale_bar_size = 1000,
  pad_left = 500,
  pad_right = 0,
  binsize = 1,
  bin_FUN = mean,
  .expand_ranges = list(
    "PROseq" = FALSE,
    "ATACDHS" = TRUE,
    "ChIPseq" = TRUE)
)

# D - geko RB
geko.gr <- genes.gr[which(genes.gr$tx_name == "geko-RB")]

# moving end of gene in (to condense plot)
end(geko.gr) <- end(geko.gr) - 3000

# Plotting (see file browserPlotR.R for more detail)
S6D <- browser_plotter.jj(
  geko.gr,
  list("PROseq" = PROseq_normed_merged.lst,
        "ATACDHS" = ATACDHS_normed_merged.lst,
        "ChIPseq" = ChIPnew_merged.lst),

```

```

labs = c("LACZ", "GAF", "BAP", "NURF", "N+B",
         "LACZ", "GAF", "BAP", "NURF", "N+B",
         "LACZGAFChIP", "GAFGAFChIP", "BAPGAFChIP"),
scale_bar_size = 1000,
pad_left = 500,
pad_right = 0,
binsize = 1,
bin_FUN = mean,
.expand_ranges = list(
  "PROseq" = FALSE,
  "ATACDHS" = TRUE,
  "ChIPseq" = TRUE)
)

# E - out-RA
out.gr <- genes.gr[which(genes.gr$tx_name == "out-RA")]

# moving end of gene in (to condense plot)
start(out.gr) <- start(out.gr) + 5000

# Plotting (see file browserPlotR.R for more detail)
S6E <- browser_plotter.jj(
  out.gr,
  list("PROseq" = PROseq_normed_merged.lst,
       "ATACDHS" = ATACDHS_normed_merged.lst,
       "ChIPseq" = ChIPnew_merged.lst),
  labs = c("LACZ", "GAF", "BAP", "NURF", "N+B",
           "LACZ", "GAF", "BAP", "NURF", "N+B",
           "LACZGAFChIP", "GAFGAFChIP", "BAPGAFChIP"),
  scale_bar_size = 1000,
  pad_left = 0,
  pad_right = 500,
  binsize = 1,
  bin_FUN = mean,
  .expand_ranges = list(
    "PROseq" = FALSE,
    "ATACDHS" = TRUE,
    "ChIPseq" = TRUE)
)

S6_all <- grid.arrange(S6A, S6B, S6C, S6D, S6E, ncol = 3, nrow = 2)

pdf("S6.pdf")
grid.draw(S6_all)
dev.off()

```

Figure S7

- A - ATAC-seq at BAP-dependent promoters
- B - PRO-seq at BAP-dependent promoters (GAF/BAP/LACZ)
- C - PRO-seq at BAP-dependent promoters (NURF/NURFBAP/LACZ)
- D - GAF PR l2FC vs NURF+BAP l2FC

```

# A - ATAC-seq at BAP-dependent promoters
# Converting results object to df
PR_BAPres.df <- as.data.frame(PR_BAP.res)
PR_BAPres.df$tx_name <- row.names(PR_BAPres.df)

# Getting list of BAP-dependent promoters (FDR 0.01)
bapDepPR.gr <-
  genes.gr[which(
    genes.gr$tx_name %in%
      (PR_BAPres.df %>%
        filter(padj < 0.01 & log2FoldChange < 0))$tx_name
  )]

# Getting counts matrix of ATAC data
bapDepPR_ATACDHSSub.mat <-
  metaSubsample(
    dataset.gr = ATACDHS_normed_merged.lst,
    regions.gr = promoters(bapDepPR.gr, 500, 500),
    binsize = 1,
    first.output.xval = -500,
    expand_ranges = TRUE
  )

# Factoring sample.name column
bapDepPR_ATACDHSSub.mat$sample.name <-
  factor(
    bapDepPR_ATACDHSSub.mat$sample.name,
    levels = names(ATACDHS_normed_merged.lst)
  )

# Plotting
S7A <- ggMetaplot(bapDepPR_ATACDHSSub.mat) +
  gg_color_scale_RNAi_ATACDHS +
  gg_fill_scale_RNAi_ATACDHS +
  scale_x_continuous(
    limits = c(-500, 500),
    expand = c(0,0)
  ) +
  scale_y_continuous(
    limits = c(0, 200),
    expand = c(0,0)
  ) +
  ggtitle("ATAC < 120 bp")

# B - PRO-seq at BAP-dependent promoters (GAF/BAP/LACZ)
# C - PRO-seq at BAP-dependent promoters (NURF/NURFBAP/LACZ)
# make one plot, duplicate in figure,
# and delete lines in illustrator

# Getting counts matrixs
bapDepPR_proSeqSub.mat <-
  metaSubsample(
    dataset.gr = PROseq_normed_merged.lst,

```

```

        regions.gr = promoters(bapDepPR.gr, 0, 150),
        binsize = 2,
        first.output.xval = 0,
        expand_ranges = FALSE
    )

bapDepGB_proSeqSub.mat <-
  metaSubsample(
    dataset.gr = PROseq_normed_merged.lst,
    regions.gr = promoters(bapDepPR.gr, 0, 1500),
    binsize = 20,
    first.output.xval = 0,
    expand_ranges = FALSE
  )

# Filling sample.name column and factoring
bapDepPR_proSeqSub.mat$sample.name <-
  factor(
    bapDepPR_proSeqSub.mat$sample.name,
    levels = names(PROseq_normed_merged.lst)
  )

bapDepGB_proSeqSub.mat$sample.name <-
  factor(
    bapDepGB_proSeqSub.mat$sample.name,
    levels = names(PROseq_normed_merged.lst)
  )

# Plotting promoter
S7Ba <- ggMetaplot(bapDepPR_proSeqSub.mat)+
  gg_color_scale_RNAi_PROseq +
  gg_fill_scale_RNAi_PROseq +
  scale_x_continuous(
    limits = c(0, 150),
    expand = c(0,0),
    breaks = c(0, 50, 100, 150)
  ) +
  scale_y_continuous(
    breaks = c(0, 50, 100)
  ) +
  coord_cartesian(
    ylim = c(0, 100),
    expand = 0,
  )+
  ggtitle("PRO-seq")

# Plotting gene body
S7Bb <- ggMetaplot(bapDepGB_proSeqSub.mat)+
  gg_color_scale_RNAi_PROseq +
  gg_fill_scale_RNAi_PROseq +

```

```

scale_y_continuous(
  breaks = c(0, 0.25, 0.5, 0.75)
)+
scale_x_continuous(
  breaks = c(seq(150, 1500, length.out = 4))
)+
coord_cartesian(
  xlim = c(150, 1500),
  ylim = c(0, 0.75),
  expand = 0,
) +
ylab(NULL)

# D - GAF PR l2FC vs NURF+BAP l2FC
S7D <- fc_corr.jj(PR_GAF.res, PR_NURFBAP.res)+
  coord_cartesian(
    xlim = c(-8,8), ylim = c(-8,8), expand = FALSE
  )+
  xlab("GAF l2FC vs. LACZ PR")+
  ylab("NURF+BAP l2FC vs. LACZ PR")

S7_all <- (S7A + S7Ba + S7Bb) / (S7Ba + S7Bb + S7D) +
  plot_layout(guides = "collect")+
  plot_annotation(tag_levels = 'A') &
  theme(plot.tag = element_text(
    family = "Helvetica", face = "bold", size = 12))

```

Figure 3

A - GB PRO-seq l2FC vs RNA-seq l2FC NURF-RNAi
 B - ATAC-MN metaprofile

```

# A - RNA-seq vs PRO-seq GB scatter plot
NURF_PRO_RNA.df <-
  data.frame(
    tx_name = row.names(as.data.frame(GB_NURF.res)),
    PRO_l2FC = as.data.frame(GB_NURF.res)$log2FoldChange,
    PRO_padj = as.data.frame(GB_NURF.res)$padj,
    RNA_l2FC = as.data.frame(RNA_NURF.res)$log2FoldChange,
    RNA_padj = as.data.frame(RNA_NURF.res)$padj
  )

# Eliminating non-significant rows
NURF_PRO_RNA.df <- drop_na(NURF_PRO_RNA.df)
NURF_PRO_RNA.df <-
  NURF_PRO_RNA.df[
    NURF_PRO_RNA.df$PRO_padj < 0.1 & NURF_PRO_RNA.df$RNA_padj < 0.1,
  ]

# Counting number points in each quadrant
num_top_left <-
  nrow(NURF_PRO_RNA.df[

```

```

    NURF_PRO_RNA.df$PRO_12FC < 0 & NURF_PRO_RNA.df$RNA_12FC > 0,
  ])
num_top_right <-
  nrow(NURF_PRO_RNA.df[
    NURF_PRO_RNA.df$PRO_12FC > 0 & NURF_PRO_RNA.df$RNA_12FC > 0,
  ])
num_bottom_right <-
  nrow(NURF_PRO_RNA.df[
    NURF_PRO_RNA.df$PRO_12FC > 0 & NURF_PRO_RNA.df$RNA_12FC < 0,
  ])
num_bottom_left <-
  nrow(NURF_PRO_RNA.df[
    NURF_PRO_RNA.df$PRO_12FC < 0 & NURF_PRO_RNA.df$RNA_12FC < 0,
  ])

# Creating annotation df for adding numbers to plot
annotations <- data.frame(
  xpos = c(-3.5, -3.5, 3.5, 3.5),
  ypos = c(-0.5, 0.5, -0.5, 0.5),
  annotateText = c(
    num_bottom_left,
    num_top_left,
    num_bottom_right,
    num_top_right))

F3A <- ggplot(NURF_PRO_RNA.df, aes(x = PRO_12FC, y = RNA_12FC)) +
  geom_point(size = 0.5, alpha = 0.75) +
  ggtheme.jj() +
  geom_text(
    data=annotations,
    aes(x=xpos, y=ypos, label=annotateText)) +
  geom_vline(xintercept = 0, linetype = "dashed", alpha = 0.75) +
  geom_hline(yintercept = 0, linetype = "dashed", alpha = 0.75) +
  coord_cartesian(xlim = c(0, 4), ylim = c(-4, 4), expand = FALSE) +
  ggtitle("PRO-seq GB vs. RNA-seq")

# B - ATAC-MN metaprofile
# Getting list of NURF-dependent promoters (FDR 0.01)
PR_NURFres.df <- as.data.frame(PR_NURF.res)
PR_NURFres.df$tx_name <- row.names(PR_NURFres.df)
nurfDepPR.gr <-
  genes.gr[which(
    genes.gr$tx_name %in%
      (PR_NURFres.df %>%
        filter(padj < 0.01 & log2FoldChange > 0))$tx_name
  )]

# Getting counts matrix of ATAC data
nurfDepPR_ATACMNSub.mat <-
  metaSubsample(
    dataset.gr = ATACMN_normed_merged.lst,
    regions.gr = promoters(nurfDepPR.gr, 500, 500),
    binsize = 5,

```

```

        first.output.xval = -500,
        expand_ranges = TRUE
    )

# Factoring sample.name column
nurfDepPR_ATACMNSub.mat$sample.name <-
    factor(
        nurfDepPR_ATACMNSub.mat$sample.name,
        levels = names(ATACMN_normed_merged.lst)
    )

# Plotting
F3B <- ggMetaplot(nurfDepPR_ATACMNSub.mat) +
    gg_color_scale_RNAi_ATACMN +
    gg_fill_scale_RNAi_ATACMN +
    scale_x_continuous(
        limits = c(-500, 500),
        expand = c(0,0)
    ) +
    scale_y_continuous(
        limits = c(0, .8),
        expand = c(0,0)
    ) +
    ggtitle("ATAC 130-200 bp")

# Laying out panels
F3_layout <- "
ABB
"

F3_all <- F3A + F3B +
    plot_layout(guides = "collect", design = F3_layout) +
    plot_annotation(tag_levels = 'A') &
    theme(plot.tag = element_text(
        family = "Helvetica", face = "bold", size = 12))

```

Figure S8

- A - MA plot of GAF vs. PBAP ATAC peaks
- B - Venn diagram of GAF and PBAP promoter intersections
- C - GAF ChIP-seq of three classes
- D - ATAC-seq metaprofiles of three classes
- E - PRO-seq pause profiles of three classes
- F - motif analysis

```

# A - MA plot of GAF vs. PBAP ATAC peaks
# Getting counts table
ATAC_peaks.df <- data.frame(ATAC_peaks.tbl, row.names = "peak_name")
ATAC_peaks_col_data.df <- data.frame(
    row.names = colnames(ATAC_peaks.df),
    RNAi = c(
        "BAP170",

```



```

    "BAP170",
    "GAF",
    "GAF",
    "LACZ",
    "LACZ",
    "NURFBAP",
    "NURFBAP",
    "NURF301",
    "NURF301"
  )
)

# Running DESeq2
ATAC_peaks_GAFvsBAP.res <-
  subset_DESeq.jj(ATAC_peaks.df,
                  ATAC_peaks_col_data.df,
                  scale_facts_ATACDHS.dbl,
                  c(3,4,1,2),
                  c("RNAi", "GAF", "BAP170"),
                  ~RNAi)

# Plotting
S8A <- maplot.jj(ATAC_peaks_GAFvsBAP.res)+
  scale_y_continuous(limits = c(-10, 10),
                    breaks = c(-10,-5,0,5,10),
                    expand= c(0,0))

# E - Venn diagram of GAF and PBAP promoter intersections
# Getting lists of GAF and BAP dependent pause genes
bapDepGenes.chr <- rownames(
  PR_BAP.res[which(PR_BAP.res$padj < 0.01 & PR_BAP.res$log2FoldChange < 0), ])
gafDepGenes.chr <- rownames(
  PR_GAF.res[which(PR_GAF.res$padj < 0.01 & PR_GAF.res$log2FoldChange < 0), ])

# Defining 3 sets of genes:
# GAF-independent but BAP-dependent
gafIndBapDep.gr <- genes.gr[
  genes.gr$tx_name %in% bapDepGenes.chr & !genes.gr$tx_name %in% gafDepGenes.chr,
]
# GAF- and BAP-dependent
gafDepBapDep.gr <- genes.gr[
  genes.gr$tx_name %in% gafDepGenes.chr & genes.gr$tx_name %in% bapDepGenes.chr
]
# GAF-dependent but BAP-independent
gafDepBapInd.gr <- genes.gr[
  genes.gr$tx_name %in% gafDepGenes.chr & !genes.gr$tx_name %in% bapDepGenes.chr,
]

# Plotting Euler diagram (area-proportional Venn diagram)
S8B <- plot(euler(c("GAF" = length(gafDepBapInd.gr),
                  "BAP" = length(gafIndBapDep.gr),

```

```

                                "GAF&BAP" = length(gafDepBapDep.gr)),
                                shape = "ellipse"),
                                quantities = TRUE)

# H - GAF ChIP-seq of three classes
# Getting signal DF
gafBapDepSorted_GAFChIP.df <-
  data.frame(
    "x" = rep(seq(-497.5, 497.5, 5), 3),
    "signal" = c(
      colMeans(getCountsByPositions(
        ChIP.lst$GAF_ChIPseq,
        promoters(gafDepBapInd.gr, 500, 500),
        binsize = 5,
        expand_ranges = TRUE
      )),
      colMeans(getCountsByPositions(
        ChIP.lst$GAF_ChIPseq,
        promoters(gafDepBapDep.gr, 500, 500),
        binsize = 5,
        expand_ranges = TRUE
      )),
      colMeans(getCountsByPositions(
        ChIP.lst$GAF_ChIPseq,
        promoters(gafIndBapDep.gr, 500, 500),
        binsize = 5,
        expand_ranges = TRUE
      ))
    ),
    "class" = c(
      rep("gafDepBapInd", 200),
      rep("gafDepBapDep", 200),
      rep("gafIndBapDep", 200)
    )
  )

# Plotting
S8C <- ggplot(gafBapDepSorted_GAFChIP.df,
              aes(x = x, y = signal, color = class))+
  geom_line()+
  ggtheme.jj()+
  scale_x_continuous(expand = c(0,0))+
  scale_y_continuous(limits = c(0, 75), breaks = c(0, 25, 50, 75), expand = c(0,0))+
  ylab("Mean")+
  xlab("Distance from TSS")+
  ggtitle("GAF ChIP-seq")

# D - ATAC-seq metaprofiles of three classes
# Getting counts matrix of ATAC data
gafIndBapDep_ATACDHSSub.mat <-
  metaSubsample(
    dataset.gr = ATACDHS_normed_merged.lst,
    regions.gr = promoters(gafIndBapDep.gr, 500, 500),

```

```

    binsize = 1,
    first.output.xval = -500,
    expand_ranges = TRUE
  )

gafDepBapDep_ATACDHSSub.mat <-
  metaSubsample(
    dataset.gr = ATACDHS_normed_merged.lst,
    regions.gr = promoters(gafDepBapDep.gr, 500, 500),
    binsize = 1,
    first.output.xval = -500,
    expand_ranges = TRUE
  )

gafDepBapInd_ATACDHSSub.mat <-
  metaSubsample(
    dataset.gr = ATACDHS_normed_merged.lst,
    regions.gr = promoters(gafDepBapInd.gr, 500, 500),
    binsize = 1,
    first.output.xval = -500,
    expand_ranges = TRUE
  )

# Factoring sample.name column
gafIndBapDep_ATACDHSSub.mat$sample.name <-
  factor(
    gafIndBapDep_ATACDHSSub.mat$sample.name,
    levels = names(ATACDHS_normed_merged.lst)
  )

gafDepBapDep_ATACDHSSub.mat$sample.name <-
  factor(
    gafDepBapDep_ATACDHSSub.mat$sample.name,
    levels = names(ATACDHS_normed_merged.lst)
  )

gafDepBapInd_ATACDHSSub.mat$sample.name <-
  factor(
    gafDepBapInd_ATACDHSSub.mat$sample.name,
    levels = names(ATACDHS_normed_merged.lst)
  )

# Plotting
S8Da <- ggMetaplot(gafDepBapInd_ATACDHSSub.mat) +
  gg_color_scale_RNAi_ATACDHS +
  gg_fill_scale_RNAi_ATACDHS +
  scale_x_continuous(
    limits = c(-500, 500),
    expand = c(0,0)
  ) +
  scale_y_continuous(
    limits = c(0, 200),
    expand = c(0,0)
  )

```

```

    )+
    ggtitle("ATAC < 120 bp")

S8Db <- ggMetaplot(gafDepBapDep_ATACDHSSub.mat) +
  gg_color_scale_RNAi_ATACDHS +
  gg_fill_scale_RNAi_ATACDHS +
  scale_x_continuous(
    limits = c(-500, 500),
    expand = c(0,0)
  ) +
  scale_y_continuous(
    limits = c(0, 200),
    expand = c(0,0)
  )+
  ggtitle("ATAC < 120 bp")

S8Dc <- ggMetaplot(gafIndBapDep_ATACDHSSub.mat) +
  gg_color_scale_RNAi_ATACDHS +
  gg_fill_scale_RNAi_ATACDHS +
  scale_x_continuous(
    limits = c(-500, 500),
    expand = c(0,0)
  ) +
  scale_y_continuous(
    limits = c(0, 200),
    expand = c(0,0)
  )+
  ggtitle("ATAC < 120 bp")

# E - PRO-seq pause profiles of three classes
## Getting counts matrix
gafIndBapDep_proSeqSubPR.mat <-
  metaSubsample(
    dataset.gr = PROseq_normed_merged.lst,
    regions.gr = promoters(gafIndBapDep.gr, 0, 150),
    binsize = 2,
    first.output.xval = 0,
    expand_ranges = FALSE
  )

gafDepBapDep_proSeqSubPR.mat <-
  metaSubsample(
    dataset.gr = PROseq_normed_merged.lst,
    regions.gr = promoters(gafDepBapDep.gr, 0, 150),
    binsize = 2,
    first.output.xval = 0,
    expand_ranges = FALSE
  )

gafDepBapInd_proSeqSubPR.mat <-
  metaSubsample(
    dataset.gr = PROseq_normed_merged.lst,

```

```

        regions.gr = promoters(gafDepBapInd.gr, 0, 150),
        binsize = 2,
        first.output.xval = 0,
        expand_ranges = FALSE
    )

# Filling sample.name column and factoring
gafIndBapDep_proSeqSubPR.mat$sample.name <-
    factor(
        gafIndBapDep_proSeqSubPR.mat$sample.name,
        levels = names(PROseq_normed_merged.lst)
    )

gafDepBapDep_proSeqSubPR.mat$sample.name <-
    factor(
        gafDepBapDep_proSeqSubPR.mat$sample.name,
        levels = names(PROseq_normed_merged.lst)
    )

gafDepBapInd_proSeqSubPR.mat$sample.name <-
    factor(
        gafDepBapInd_proSeqSubPR.mat$sample.name,
        levels = names(PROseq_normed_merged.lst)
    )

# Plotting promoter
S8Ea <- ggMetaplot(gafDepBapInd_proSeqSubPR.mat)+
    gg_color_scale_RNAi_PROseq +
    gg_fill_scale_RNAi_PROseq +
    scale_x_continuous(
        limits = c(0, 150),
        expand = c(0,0),
        breaks = c(0, 50, 100, 150)
    ) +
    scale_y_continuous(
        breaks = c(0, 25, 50, 75)
    ) +
    coord_cartesian(
        ylim = c(0, 75),
        expand = 0,
    )+
    ggtitle("PRO-seq")

S8Eb <- ggMetaplot(gafDepBapDep_proSeqSubPR.mat)+
    gg_color_scale_RNAi_PROseq +
    gg_fill_scale_RNAi_PROseq +
    scale_x_continuous(
        limits = c(0, 150),
        expand = c(0,0),
        breaks = c(0, 50, 100, 150)
    ) +

```

```

    scale_y_continuous(
      breaks = c(0, 25, 50, 75)
    ) +
    coord_cartesian(
      ylim = c(0, 75),
      expand = 0,
    ) +
    ggtitle("PR0-seq")

S8Ec <- ggMetaplot(gafIndBapDep_proSeqSubPR.mat) +
  gg_color_scale_RNAi_PROseq +
  gg_fill_scale_RNAi_PROseq +
  scale_x_continuous(
    limits = c(0, 150),
    expand = c(0,0),
    breaks = c(0, 50, 100, 150)
  ) +
  scale_y_continuous(
    breaks = c(0,25, 50, 75)
  ) +
  coord_cartesian(
    ylim = c(0, 75),
    expand = 0,
  ) +
  ggtitle("PR0-seq")

# I - motif analysis
# Saving bed files of promoters. I'll take these, use
# bedtools getfasta -s to get stranded sequences, and
# then use DREME to search gafIndBapDep and gafDepBapInd
# vs. gafDepBapDep using DREME, e < 0.001.
# Then input found motifs in TomTom
export.bed(unique(
  promoters(gafIndBapDep.gr, 500, 0)), "bed/gafIndBapDep.bed")
export.bed(unique(
  promoters(gafDepBapInd.gr, 500, 0)), "bed/gafDepBapInd.bed")
export.bed(unique(
  promoters(gafDepBapDep.gr, 500, 0)), "bed/gafDepBapDep.bed")

# Combining all panels
S8_layout <- "
ABBC
#DEF
#GHI
"

S8_all <- S8A + S8B + S8C + S8Da + S8Db + S8Dc + S8Ea + S8Eb + S8Ec +
  plot_layout(guides = "collect", design = S9_layout) +
  plot_annotation(tag_levels = 'A') &
  theme(plot.tag = element_text(
    family = "Helvetica", face = "bold", size = 12))

```

Figure S9

- A - HM of GAF/NURF CUT&RUN at all promoters sorted by GAF signal
- B - NURF GBupRNAup and GBupRNAdown sorted PI distributions
- C - NURF GBupRNAup and GBupRNAdown sorted LACZ RNA-seq distributions
- D - NURF GBupRNAup and GBupRNAdown sorted ATACDHS metaprofiles
- E - NURF GBupRNAup and GBupRNAdown sorted ATACMN metaprofiles
- F - NURF GBupRNAup and GBupRNAdown sorted PRO-seq metaprofiles

```
# A - HM of GAF/NURF CUT&RUN at all promoters sorted by GAF signal
# Getting total GAF signal from TSS-500 to TSS in new genes.gr object
genes_gafCaRSorted.gr <- genes.gr
genes_gafCaRSorted.gr$GAF_CaR_signal <-
  getCountsByRegions(
    CaR.lst$GAF_CUTandRUN,
    promoters(genes_gafCaRSorted.gr, 500, 0),
    expand_ranges = TRUE
  )

# Sorting by GAF CUT&RUN signal
genes_gafCaRSorted.gr <-
  sort(genes_gafCaRSorted.gr, by = ~GAF_CaR_signal, decreasing = TRUE)

# Taking only top quartile
genes_gafCaRSorted.gr <-
  genes_gafCaRSorted.gr[
    genes_gafCaRSorted.gr$GAF_CaR_signal > quantile(
      genes_gafCaRSorted.gr$GAF_CaR_signal)[[4]],]

# Getting signal matrix
genes_gafCaRSorted_CaR.mat <- cbind(
  getCountsByPositions(
    dataset.gr = CaR.lst$GAF_CUTandRUN,
    regions.gr = promoters(genes_gafCaRSorted.gr, 500, 500),
    binsize = 10,
    expand_ranges = TRUE
  ),
  getCountsByPositions(
    dataset.gr = CaR.lst$NURF_CUTandRUN,
    regions.gr = promoters(genes_gafCaRSorted.gr, 500, 500),
    binsize = 10,
    expand_ranges = TRUE
  )
)

# Plotting Heatmap
col_fun = colorRamp2(c(0, 10, 20), viridis(3))
S9A <- Heatmap(
  genes_gafCaRSorted_CaR.mat,
  cluster_columns = F,
  cluster_rows = F,
  col = col_fun,
  use_raster = F
)
```

```

# B - NURF GBupRNAup and GBupRNAdown sorted PI distributions
# Getting lists of genes
NURFupRNAup.gr <-
  genes.gr[
    genes.gr$tx_name %in%
      NURF_PRO_RNA.df[
        NURF_PRO_RNA.df$PRO_12FC > 0 & NURF_PRO_RNA.df$RNA_12FC > 0,
      ]$tx_name
  ]

NURFupRNAdown.gr <-
  genes.gr[
    genes.gr$tx_name %in%
      NURF_PRO_RNA.df[
        NURF_PRO_RNA.df$PRO_12FC > 0 & NURF_PRO_RNA.df$RNA_12FC < 0,
      ]$tx_name
  ]

# Adding gene length
NURFupRNAup.gr$length <- lengths(NURFupRNAup.gr)
NURFupRNAdown.gr$length <- lengths(NURFupRNAdown.gr)

# Adding pause LACZ PRO-seq signal
NURFupRNAup.gr$pause_signal <-
  getCountsByRegions(
    PROseq_normed_merged.lst$LACZ_PROseq,
    genebodies(NURFupRNAup.gr, -50, 100, fix.end = "start")
  )

NURFupRNAdown.gr$pause_signal <-
  getCountsByRegions(
    PROseq_normed_merged.lst$LACZ_PROseq,
    genebodies(NURFupRNAdown.gr, -50, 100, fix.end = "start")
  )

# Adding GB LACZ PRO-seq signal
NURFupRNAup.gr$GB_signal <-
  getCountsByRegions(
    PROseq_normed_merged.lst$LACZ_PROseq,
    genebodies(NURFupRNAup.gr, 200, -200)
  )

NURFupRNAdown.gr$GB_signal <-
  getCountsByRegions(
    PROseq_normed_merged.lst$LACZ_PROseq,
    genebodies(NURFupRNAdown.gr, 200, 200)
  )

# Adding norm factor for length normalizing GB signal
NURFupRNAup.gr$length_scaleFact <- 150 / NURFupRNAup.gr$length
NURFupRNAdown.gr$length_scaleFact <- 150 / NURFupRNAdown.gr$length

```



```

# Calculating PI
NURFupRNAup.gr$PI <-
  NURFupRNAup.gr$pause_signal /
  (NURFupRNAup.gr$GB_signal *
    NURFupRNAup.gr$length_scaleFact)

NURFupRNAdown.gr$PI <-
  NURFupRNAdown.gr$pause_signal /
  (NURFupRNAdown.gr$GB_signal *
    NURFupRNAdown.gr$length_scaleFact)

# Getting dataframe of PIs
PI.df <- rbind(
  data.frame(class = "NURFupRNAup", PI = NURFupRNAup.gr$PI),
  data.frame(class = "NURFupRNAdown", PI = NURFupRNAdown.gr$PI)
)

S9B <- ggplot(PI.df, aes(x = class, y = PI, fill = class))+
  geom_violin()+
  geom_boxplot(width = 0.2, fill = "white", outlier.shape = NA)+
  stat_compare_means(method = "wilcox.test")+
  ggtheme.jj()+
  scale_fill_manual(values = updown_cols.chr,
                    breaks = c("NURFupRNAup", "NURFupRNAdown"),
                    labels = c("up_up", "up_down")) +
  theme(axis.text.x = element_blank(),
        axis.ticks.x = element_blank(),
        axis.title.x = element_blank(),
        axis.line.x = element_blank())+
  ylab("log10 Pause Index")+
  scale_y_log10(
    limits = c(1e-2, 1e4),
    breaks = c(1e-2, 1e0, 1e2, 1e4),
    labels = c("-2", "0", "2", "4"),
    expand = c(0,0)
  )

# C - NURF GBupRNAup and GBupRNAdown sorted LACZ RNA-seq distributions
# Adding LACZ RNA-seq signal
NURFupRNAup.gr$RNA_signal <-
  getCountsByRegions(
    RNAseq_normed_merged.lst$LACZ_RNAseq,
    genebodies(NURFupRNAup.gr, -1000, 0, fix.start = "end")
  )

NURFupRNAdown.gr$RNA_signal <-
  getCountsByRegions(
    RNAseq_normed_merged.lst$LACZ_RNAseq,
    genebodies(NURFupRNAdown.gr, -1000, 0, fix.start = "end")
  )

RNA.df <- rbind(

```

```

data.frame(class = "NURFupRNAup", RNA = NURFupRNAup.gr$RNA_signal),
data.frame(class = "NURFupRNAdown", RNA = NURFupRNAdown.gr$RNA_signal)
)

S9C <- ggplot(RNA.df, aes(x = class, y = RNA, fill = class))+
  geom_violin()+
  geom_boxplot(width = 0.2, fill = "white", outlier.shape = NA)+
  stat_compare_means(method = "wilcox.test")+
  ggtheme.jj()+
  scale_fill_manual(values = updown_cols.chr,
                    breaks = c("NURFupRNAup", "NURFupRNAdown"),
                    labels = c("up_up", "up_down")) +
  theme(axis.text.x = element_blank(),
        axis.ticks.x = element_blank(),
        axis.title.x = element_blank(),
        axis.line.x = element_blank())+
  ylab("log10 RNA-seq")+
  scale_y_log10(
    limits = c(1e0, 1e6),
    breaks = c(1e0, 1e2, 1e4, 1e6),
    labels = c("0", "2", "4", "6"),
    expand = c(0,0)
  )

# D - NURF GBupRNAup and GBupRNAdown sorted ATACDHS metaprofiles
NURFupRNAupATACDHS.df <-
  metaSubsample(
    dataset.gr = ATACDHS_normed_merged.lst,
    regions.gr = promoters(NURFupRNAup.gr, 500, 500),
    binsize = 1,
    first.output.xval = -500,
    expand_ranges = TRUE
  )

NURFupRNAupATACDHS.df$class <- "up_up"

NURFupRNAdownATACDHS.df <-
  metaSubsample(
    dataset.gr = ATACDHS_normed_merged.lst,
    regions.gr = promoters(NURFupRNAdown.gr, 500, 500),
    binsize = 1,
    first.output.xval = -500,
    expand_ranges = TRUE
  )

NURFupRNAdownATACDHS.df$class <- "up_down"

nurfsorted_ATACDHS.df <- rbind(NURFupRNAupATACDHS.df,
                              NURFupRNAdownATACDHS.df)

S9D <- ggMetaplot(nurfsorted_ATACDHS.df) +
  gg_color_scale_RNAi_ATACDHS +
  gg_fill_scale_RNAi_ATACDHS +

```

```

    scale_x_continuous(
      limits = c(-500, 500),
      expand = c(0,0)
    ) +
    scale_y_continuous(
      limits = c(0, 120),
      expand = c(0,0)
    ) +
    ggtitle("ATAC < 120 bp") +
    facet_grid(~class)

# E - NURF GBupRNAup and GBupRNAdown sorted ATACMN metaprofiles
NURFupRNAupATACMN.df <-
  metaSubsample(
    dataset.gr = ATACMN_normed_merged.lst,
    regions.gr = promoters(NURFupRNAup.gr, 500, 500),
    binsize = 5,
    first.output.xval = -500,
    expand_ranges = TRUE
  )

NURFupRNAupATACMN.df$class <- "up_up"

NURFupRNAdownATACMN.df <-
  metaSubsample(
    dataset.gr = ATACMN_normed_merged.lst,
    regions.gr = promoters(NURFupRNAdown.gr, 500, 500),
    binsize = 5,
    first.output.xval = -500,
    expand_ranges = TRUE
  )

NURFupRNAdownATACMN.df$class <- "up_down"

nurfSorted_ATACMN.df <- rbind(NURFupRNAupATACMN.df,
                              NURFupRNAdownATACMN.df)

S9E <- ggMetaplot(nurfSorted_ATACMN.df) +
  gg_color_scale_RNAi_ATACDHS +
  gg_fill_scale_RNAi_ATACDHS +
  scale_x_continuous(
    limits = c(-500, 500),
    expand = c(0,0)
  ) +
  scale_y_continuous(
    limits = c(0, 1.0),
    expand = c(0,0)
  ) +
  ggtitle("ATAC 130-200 bp") +
  facet_grid(~class)

# F - NURF GBupRNAup and GBupRNAdown sorted PRO-seq metaprofiles

```

```

NURFupRNAup_PR.mat <-
  metaSubsample(
    dataset.gr = PROseq_normed_merged.lst,
    regions.gr = promoters(NURFupRNAup.gr, 0, 150),
    binsize = 2,
    first.output.xval = 0,
    expand_ranges = FALSE
  )

NURFupRNAup_PR.mat$class <- "up_up"

NURFupRNAdown_PR.mat <-
  metaSubsample(
    dataset.gr = PROseq_normed_merged.lst,
    regions.gr = promoters(NURFupRNAdown.gr, 0, 150),
    binsize = 2,
    first.output.xval = 0,
    expand_ranges = FALSE
  )

NURFupRNAdown_PR.mat$class <- "up_down"

NURFupRNAup_GB.mat <-
  metaSubsample(
    dataset.gr = PROseq_normed_merged.lst,
    regions.gr = promoters(NURFupRNAup.gr, 0, 1500),
    binsize = 20,
    first.output.xval = 0,
    expand_ranges = FALSE
  )

NURFupRNAup_GB.mat$class <- "up_up"

NURFupRNAdown_GB.mat <-
  metaSubsample(
    dataset.gr = PROseq_normed_merged.lst,
    regions.gr = promoters(NURFupRNAdown.gr, 0, 1500),
    binsize = 20,
    first.output.xval = 0,
    expand_ranges = FALSE
  )

NURFupRNAdown_GB.mat$class <- "up_down"

# Filling sample.name column and factoring
NURFupRNAup_PR.mat$sample.name <-
  factor(
    NURFupRNAup_PR.mat$sample.name,
    levels = names(PROseq_normed_merged.lst)
  )

NURFupRNAdown_PR.mat$sample.name <-
  factor(
    NURFupRNAdown_PR.mat$sample.name,
    levels = names(PROseq_normed_merged.lst)
  )

```

```

NURFupRNAup_GB.mat$sample.name <-
  factor(
    NURFupRNAup_GB.mat$sample.name,
    levels = names(PROseq_normed_merged.lst)
  )

NURFupRNAdown_GB.mat$sample.name <-
  factor(
    NURFupRNAdown_GB.mat$sample.name,
    levels = names(PROseq_normed_merged.lst)
  )

nurfsorted_PR.mat <- rbind(
  NURFupRNAup_PR.mat, NURFupRNAdown_PR.mat
)

nurfsorted_GB.mat <- rbind(
  NURFupRNAup_GB.mat, NURFupRNAdown_GB.mat
)

# Plotting promoter
S9Fa <- ggMetaplot(nurfsorted_PR.mat)+
  gg_color_scale_RNAi_PROseq +
  gg_fill_scale_RNAi_PROseq +
  scale_x_continuous(
    limits = c(0, 150),
    expand = c(0,0),
    breaks = c(0, 50, 100, 150)
  ) +
  scale_y_continuous(
    breaks = c(0, 50, 100, 150)
  ) +
  coord_cartesian(
    ylim = c(0, 150),
    expand = FALSE,
  )+
  ggtitle("PRO-seq")+
  facet_grid(class~.)

# Plotting gene body
S9Fb <- ggMetaplot(nurfsorted_GB.mat)+
  gg_color_scale_RNAi_PROseq +
  gg_fill_scale_RNAi_PROseq +
  scale_y_continuous(
    breaks = c(0, 1, 2, 3)
  )+
  scale_x_continuous(
    breaks = c(seq(150, 1500, length.out = 4))
  )+
  coord_cartesian(
    xlim = c(150, 1500),
    ylim = c(0, 3),
    expand = 0,
  )

```

```

    ) +
    ylab(NULL)+
    facet_grid(class~.)

S9_BCD <- (plot_spacer() | S9B | S9C | S9D) +
  plot_layout(guides = "collect")+
  plot_annotation(tag_levels = 'A') &
  theme(plot.tag = element_text(
    family = "Helvetica", face = "bold", size = 12))

S9EF <- (S9E / S9F) + plot_layout(guides = "collect")+
  plot_annotation(tag_levels = 'A') &
  theme(plot.tag = element_text(
    family = "Helvetica", face = "bold", size = 12))

S9G <- (S9Ga | S9Gb) + plot_layout(guides = "collect")+
  plot_annotation(tag_levels = 'A') &
  theme(plot.tag = element_text(
    family = "Helvetica", face = "bold", size = 12))

```

Supplementary Code 7. Session Info

The output of this code chunk displays all the versions of packages that were used to analyze the data presented in this manuscript

```
sessionInfo()
```

```
## R version 3.6.3 (2020-02-29)
## Platform: x86_64-apple-darwin15.6.0 (64-bit)
## Running under: macOS Catalina 10.15.5
##
## Matrix products: default
## BLAS:   /Library/Frameworks/R.framework/Versions/3.6/Resources/lib/libRblas.0.dylib
## LAPACK: /Library/Frameworks/R.framework/Versions/3.6/Resources/lib/libRlapack.dylib
##
## locale:
##  [1] en_US.UTF-8/en_US.UTF-8/en_US.UTF-8/C/en_US.UTF-8/en_US.UTF-8
##
## attached base packages:
##  [1] grid      parallel  stats4     stats      graphics  grDevices  utils
##  [8] datasets  methods   base
##
## other attached packages:
##  [1] gridExtra_2.3           eulerr_6.1.0
##  [3] tiff_0.1-5             circlize_0.4.10
##  [5] ComplexHeatmap_2.0.0    plyr_1.8.6
##  [7] patchwork_1.0.1         extrafont_0.17
##  [9] BRGenomics_0.99.26      rtracklayer_1.46.0
## [11] scales_1.1.1           viridis_0.5.1
## [13] viridisLite_0.3.0      ggpubr_0.4.0
## [15] DESeq2_1.26.0          SummarizedExperiment_1.16.1
## [17] DelayedArray_0.12.2     BiocParallel_1.20.1
## [19] matrixStats_0.56.0      Biobase_2.46.0
## [21] GenomicRanges_1.38.0    GenomeInfoDb_1.22.0
## [23] IRanges_2.20.2          S4Vectors_0.24.3
## [25] BiocGenerics_0.32.0     forcats_0.5.0
## [27] stringr_1.4.0           dplyr_1.0.0
## [29] purrr_0.3.4             readr_1.3.1
## [31] tidyr_1.1.0            tibble_3.0.2
## [33] ggplot2_3.3.2          tidyverse_1.3.0
##
## loaded via a namespace (and not attached):
##  [1] readxl_1.3.1           backports_1.1.8         Hmisc_4.4-0
##  [4] splines_3.6.3          digest_0.6.25           htmltools_0.5.0
##  [7] fansi_0.4.1            magrittr_1.5            checkmate_2.0.0
## [10] memoise_1.1.0          cluster_2.1.0           openxlsx_4.1.5
## [13] Biostrings_2.54.0      annotate_1.64.0         modelr_0.1.8
## [16] extrafontdb_1.0        jpeg_0.1-8.1           colorspace_1.4-1
## [19] blob_1.2.1            rvest_0.3.5            haven_2.3.1
## [22] xfun_0.15             crayon_1.3.4           RCurl_1.98-1.2
## [25] jsonlite_1.7.0         genefilter_1.68.0       survival_3.2-3
## [28] glue_1.4.1            gtable_0.3.0           zlibbioc_1.32.0
## [31] XVector_0.26.0         GetoptLong_1.0.2       car_3.0-8
```

## [34] Rttf2pt1_1.3.8	shape_1.4.4	abind_1.4-5
## [37] DBI_1.1.0	rstatix_0.6.0	Rcpp_1.0.5
## [40] xtable_1.8-4	htmlTable_2.0.1	clue_0.3-57
## [43] foreign_0.8-76	bit_1.1-15.2	Formula_1.2-3
## [46] htmlwidgets_1.5.1	httr_1.4.1	RColorBrewer_1.1-2
## [49] acepack_1.4.1	ellipsis_0.3.1	pkgconfig_2.0.3
## [52] XML_3.99-0.3	nnet_7.3-14	dbplyr_1.4.4
## [55] locfit_1.5-9.4	tidyselect_1.1.0	rlang_0.4.7
## [58] AnnotationDbi_1.48.0	munsell_0.5.0	cellranger_1.1.0
## [61] tools_3.6.3	cli_2.0.2	generics_0.0.2
## [64] RSQLite_2.2.0	broom_0.7.0	evaluate_0.14
## [67] yaml_2.2.1	knitr_1.29	bit64_0.9-7
## [70] fs_1.4.2	zip_2.0.4	xml2_1.3.2
## [73] compiler_3.6.3	rstudioapi_0.11	curl_4.3
## [76] png_0.1-7	ggsignif_0.6.0	reprex_0.3.0
## [79] geneplotter_1.64.0	stringi_1.4.6	lattice_0.20-41
## [82] Matrix_1.2-18	vctrs_0.3.1	pillar_1.4.5
## [85] lifecycle_0.2.0	GlobalOptions_0.1.2	data.table_1.12.8
## [88] bitops_1.0-6	R6_2.4.1	latticeExtra_0.6-29
## [91] rio_0.5.16	assertthat_0.2.1	rjson_0.2.20
## [94] withr_2.2.0	GenomicAlignments_1.22.1	Rsamtools_2.2.3
## [97] GenomeInfoDbData_1.2.2	hms_0.5.3	rpart_4.1-15
## [100] rmarkdown_2.3	carData_3.0-4	lubridate_1.7.9
## [103] base64enc_0.1-3		