# CUT&RUN BMDM WT vs KO - Figures

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# Contents

Introduction	3
Load Packages	4
Figure S6A	4
Rep. 1	5
Rep. 2	5
Figure S6B	6
Figure S6C	7
Figure 6A	8
Histogram	8
Heatmap	9
Figure 6B	10
Figure 6C	11
Figure 6D	11
Figure 6E	13
Figure 6F	14
Figure 6H	15

	15
LPM	18
KC	19
$IM \ \dots $	19
Ly6C CM	20
MHC2 CM	21
MG	22

#### Introduction

Despite substantial evidence pointing to MafB as an essential regulator of core mac identity, the genes it directly controls and the mechanisms underlying this regulation remain poorly understood. To address this question, we performed cleavage under targets and release using nuclease (CUT&RUN) for MafB on BMDMs from Mafbfl/fl and Lyz2CreMafbfl/fl mice. MafB and H3K27ac CUT&RUN was performed with a CUTANA ChIC/CUT&RUN Kit (EpiCypher, 141048) according to manufacturer's instructions, with modifications. BMDMs from Mafbfl/fl or Lyz2CreMafbfl/fl mice were washed with ice-cold PBS and fixed with 0.1 % formaldehyde in PBS for 2 min at room temperature. Fixation was quenched by adding glycine (Merck, 104691000) to 0.125 M. For each CUT&RUN sample, 1 x 10^6 fixed cells and 0.5 µg antibodies were added:

IgG (EpiCypher, 13-0042)

MafB2 (Cell Signaling Technology, 41019)

H3K27ac (Thermo Fisher Scientific, MA5-23516).

CUT&RUN libraries were prepared with a CUTANA<sup>TM</sup> CUT&RUN Library Prep Kit (EpiCypher, 141001) according to manufacturer's instructions. These libraries were sequenced on an NovaSeq 6000 (Illumina) sequencer on an S4 flow cell at 10 million reads per sample.

# Load Packages

```
suppressMessages({
    library(ggplot2)
    library(colorRamp2)
    library(ComplexHeatmap)
    library(readxl)
    library(futile.logger)
    library(grid)
    library(rtracklayer)
    library(dplyr)
    library(VennDiagram)
    library(Rtoolbox)
    library(SeuratObject)
    library(Seurat)
    library(introdataviz)
}
```

## Warning: package 'colorRamp2' was built under R version 4.4.3

# Figure S6A

## [1] 14436

```
MafB2_peaks_rep1 <- read.table("MafB2_WT_R1.seacr.peaks.stringent.bed",
    header = FALSE, sep = "\t")

MafB2_peaks_rep2 <- read.table("MafB2_WT_R2.seacr.peaks.stringent.bed",
    header = FALSE, sep = "\t")

MafB2_peaks_consensus <- read.table("MafB2_WT.seacr.consensus.peak_counts.bed",
    header = FALSE, sep = "\t")

## peaks rep1
nrow(MafB2_peaks_rep1)

## [1] 10549

## peaks rep2
nrow(MafB2_peaks_rep2)

## [1] 11478

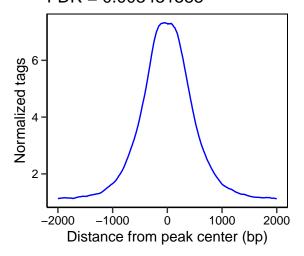
# consensus peaks
nrow(MafB2_peaks_consensus)</pre>
```

```
# shared peaks
length(which(MafB2_peaks_consensus[, 10] == 2))
## [1] 7256
```

#### Rep. 1

# Rep. 1: 10549 peaks FDR = 0.003431588

linetype = "solid"))

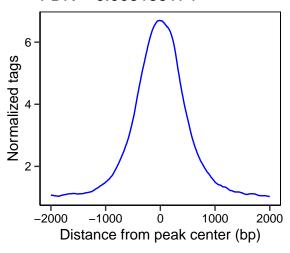


```
ggsave("Figure_S6A_Rep1.pdf", width = 3, height = 3)
```

Rep. 2

Rep. 2: 11478 peaks FDR = 0.003159174

linetype = "solid"))



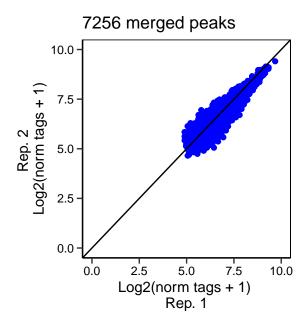
```
ggsave("Figure_S6A_Rep2.pdf.pdf", width = 3, height = 3)
```

# Figure S6B

```
samtools view -c -F 4 MafB2_WT_R1.target.markdup.sorted.bam
bedtools intersect -a MafB2_WT_R1.target.markdup.sorted.bam -b MafB2_WT_R1.seacr.peaks.stringent.bed -b
samtools view -c -F 4 MafB2_WT_R2.target.markdup.sorted.bam
bedtools intersect -a MafB2_WT_R2.target.markdup.sorted.bam -b MafB2_WT_R2.seacr.peaks.stringent.bed -b
```

## Figure S6C

```
MafB2 peaks merged <- MafB2 peaks consensus[MafB2 peaks consensus[,
    10] == 2, ]
split_scores <- strsplit(as.character(MafB2_peaks_merged$V6),</pre>
means <- sapply(split_scores, function(x) {</pre>
   if (all(is.na(x)))
       return(NA_real_)
   nums <- as.numeric(x)</pre>
   mean(nums, na.rm = TRUE)
})
MafB2_peaks_merged$V4 <- means
write.table(MafB2_peaks_merged, "MafB2_peaks_merged.bed", sep = "\t",
    quote = FALSE, row.names = FALSE, col.names = FALSE)
annotatePeaks.pl MafB2_peaks_merged.bed mm10 -d MafB2_WT_R1/ MafB2_WT_R2/ MafB2_K0_R1/ MafB2_K0_R2/ > M
MafB2_counts <- read.table("MafB2_counts.txt", header = TRUE,</pre>
    sep = "\t")
colnames(MafB2_counts) [colnames(MafB2_counts) == "PeakID..cmd.annotatePeaks.pl.MafB2_peaks_merged.bed.m
colnames(MafB2_counts)[colnames(MafB2_counts) == "MafB2_WT_R1..Tag.Count.in.given.bp..5577784.0.Total...
colnames(MafB2_counts)[colnames(MafB2_counts) == "MafB2_WT_R2..Tag.Count.in.given.bp..6205913.0.Total...
colnames(MafB2_counts)[colnames(MafB2_counts) == "MafB2_KO_R1..Tag.Count.in.given.bp..5444540.0.Total...
colnames(MafB2_counts)[colnames(MafB2_counts) == "MafB2_KO_R2..Tag.Count.in.given.bp..6810425.0.Total...
ggplot(data = MafB2_counts, aes(x = log2(MafB2_WT_R1 + 1), y = log2(MafB2_WT_R2 +
    1))) + geom point(show.legend = FALSE, colour = "blue") +
    geom_abline(slope = 1) + ggtitle("7256 merged peaks") + xlab("Log2(norm tags + 1) \nRep. 1") +
   ylab("Rep. 2 \nlog2(norm tags + 1)") + xlim(0, 10) + ylim(0,
   10) + theme_classic() + theme(axis.text.x = element_text(color = "black"),
    axis.text.y = element_text(color = "black"), axis.ticks.length = unit(0.15,
        "cm"), panel.border = element_rect(fill = NA, color = "black",
       linetype = "solid"))
```

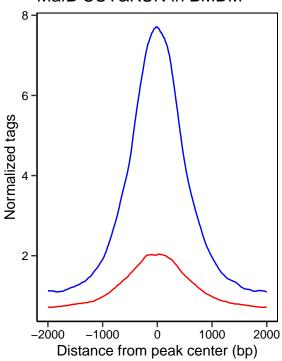


```
ggsave("Figure_S6C.pdf", width = 3, height = 3.225)
```

# Figure 6A

#### Histogram

#### MafB CUT&RUN in BMDM



MafB2\_KO\_R2\_heatmap <- MafB2\_heatmap[, 485:645]</pre>

MafB2\_KO\_heatmap <- (MafB2\_KO\_R1\_heatmap + MafB2\_KO\_R2\_heatmap)/2

```
ggsave("Figure_6A_WTvsKO_Hist.pdf", width = 3, height = 4)
```

#### Heatmap

#### $\mathbf{W}\mathbf{T}$

```
col_fun = colorRamp2(c(0, 15), c("white", "red"))

pdf("Figure_6A_heatmap_WT.pdf", width = 2, height = 4)
Heatmap(MafB2_WT_heatmap, col_fun, cluster_columns = F, cluster_rows = F,
    heatmap_legend_param = list(title = "Tags"), use_raster = TRUE,
    raster_quality = 10, column_title = "Mafbfl/fl")
```

```
## Warning: The input is a data frame-like object, convert it to a matrix.
## 'magick' package is suggested to install to give better rasterization.
## Set 'ht_opt$message = FALSE' to turn off this message.
dev.off()
## pdf
##
KO
col_fun = colorRamp2(c(0, 15), c("white", "red"))
pdf("Figure_6A_heatmap_KO.pdf", width = 2, height = 4)
Heatmap(MafB2_KO_heatmap, col_fun, cluster_columns = F, cluster_rows = F,
   heatmap_legend_param = list(title = "Tags"), use_raster = TRUE,
   raster_quality = 10, , column_title = "Lyz2CreMafbfl/fl")
## Warning: The input is a data frame-like object, convert it to a matrix.
## 'magick' package is suggested to install to give better rasterization.
## Set 'ht_opt$message = FALSE' to turn off this message.
dev.off()
## pdf
##
    2
```

# Figure 6B

# 

```
ggsave("Figure_6B.pdf", width = 3, height = 3.225)
```

# Figure 6C

```
annotatePeaks.pl MafB2_peaks.bed genome.fa -gtf genes.gtf > MafB2_peaks_annot.txt
MafB2_peaks_annot <- read.table("MafB2_peaks_annot.txt", header = TRUE,</pre>
    sep = "\t")
MafB2_peaks_annot$Annotation.not.detailed <- sub(" .*", "", MafB2_peaks_annot$Annotation)
table(MafB2_peaks_annot$Annotation.not.detailed)
##
##
                  Intergenic
                                    intron promoter-TSS
                                                                  TTS
           exon
##
            472
                         1897
                                      2423
                                                    2271
                                                                  178
```

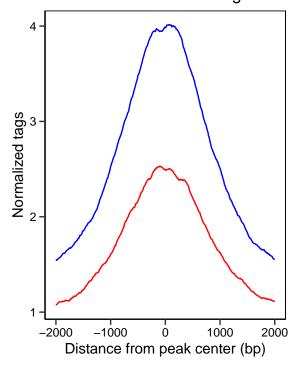
Figure 6D

```
annotatePeaks.pl MafB2_peaks.bed mm10 -size 4000 -hist 10 -d tag_directory_WT-H3K27Ac_R1/ tag_directory
```

theme\_classic() + theme(axis.text.x = element\_text(color = "black"),

## H3K27ac at MafB binding sites

linetype = "solid"))



```
ggsave("Figure_6D_Hist_H3K27Ac.pdf", height = 4, width = 3)
```

## Figure 6E

```
annotatePeaks.pl MafB2_peaks.bed mm10 -d tag_directory_WT-H3K27Ac_R1/ tag_directory_WT-H3K27Ac_R2/ tag_
MafB2_H3K27Ac_counts <- read.table("MafB2_H3K27Ac_counts.txt",</pre>
    header = TRUE, sep = "\t")
colnames(MafB2_H3K27Ac_counts)[colnames(MafB2_H3K27Ac_counts) ==
    "PeakID..cmd.annotatePeaks.pl.MafB2_peaks.bed.mm10..d.tag_directory_WT.H3K27Ac_R1..tag_directory_WT
colnames(MafB2_H3K27Ac_counts)[colnames(MafB2_H3K27Ac_counts) ==
    "tag_directory_WT.H3K27Ac_R1..Tag.Count.in.given.bp..3449599.0.Total..normalization.factor...2.90...
colnames(MafB2_H3K27Ac_counts)[colnames(MafB2_H3K27Ac_counts) ==
    "tag_directory_WT.H3K27Ac_R2..Tag.Count.in.given.bp..3954896.0.Total..normalization.factor...2.53...
colnames(MafB2_H3K27Ac_counts)[colnames(MafB2_H3K27Ac_counts) ==
    "tag_directory_KO.H3K27Ac_R1..Tag.Count.in.given.bp..3206024.0.Total..normalization.factor...3.12...
colnames(MafB2 H3K27Ac counts)[colnames(MafB2 H3K27Ac counts) ==
    "tag_directory_KO.H3K27Ac_R2..Tag.Count.in.given.bp..3641165.0.Total..normalization.factor...2.75...
MafB2_H3K27Ac_counts$H3K27Ac_WT <- rowMeans(MafB2_H3K27Ac_counts[,</pre>
    c("H3K27Ac_WT_R1", "H3K27Ac_WT_R2")])
MafB2_H3K27Ac_counts$H3K27Ac_KO <- rowMeans(MafB2_H3K27Ac_counts[,</pre>
    c("H3K27Ac_KO_R1", "H3K27Ac_KO_R2")])
MafB2_H3K27Ac_counts$1og2_H3K27Ac_WT <- log2(MafB2_H3K27Ac_counts$H3K27Ac_WT +
    1)
MafB2_H3K27Ac_counts$log2_H3K27Ac_K0 <- log2(MafB2_H3K27Ac_counts$H3K27Ac_K0 +
MafB2 H3K27Ac counts$below diag <- MafB2 H3K27Ac counts$log2 H3K27Ac KO <
    MafB2_H3K27Ac_counts$log2_H3K27Ac_WT
ggplot(data = MafB2_H3K27Ac_counts, aes(x = log2_H3K27Ac_WT,
    y = log2_H3K27Ac_K0)) + geom_point(aes(color = below_diag),
    show.legend = FALSE) + scale color manual(values = c(`FALSE` = "red",
    `TRUE` = "blue")) + geom_abline(slope = 1) + ggtitle("H3K27ac at MafB binding sites") +
    xlab("Log2(norm tags + 1) \nMafbfl/fl 1") + ylab("Lyz2CreMafbfl/fl \nLog2(norm tags + 1)") +
    xlim(0, 10) + ylim(0, 10) + theme_classic() + theme(axis.text.x = element_text(color = "black"),
    axis.text.y = element_text(color = "black"), axis.ticks.length = unit(0.15,
        "cm"), panel.border = element_rect(fill = NA, color = "black",
        linetype = "solid"))
```

# H3K27ac at MafB binding sit

```
ggsave("Figure_6E_WTvsKO_H3K27ac_Tags.pdf", height = 3.225, width = 3)
```

# Figure 6F

```
MafB2 genes <- MafB2 peaks annot [MafB2 peaks annot [, "Distance.to.TSS"] >=
    -2000 & MafB2_peaks_annot[, "Distance.to.TSS"] <= 2000, ]
MafB2_genes <- unique(MafB2_genes$Gene.Name)</pre>
MafB2_genes <- na.omit(MafB2_genes)</pre>
Mac_sign <- read_excel("Mac_sign.xlsx")</pre>
Mac_sign_genes <- Mac_sign$Gene_Symbol</pre>
# Helper function to display Venn diagram
display_venn <- function(x, ...) {</pre>
    grid.newpage()
    venn_object <- venn.diagram(x, filename = NULL, ...)</pre>
    grid.draw(venn_object)
}
list <- list(MafB2_target = MafB2_genes, Mac_sign = Mac_sign_genes)</pre>
pdf(file = "Figure_6F_Venn.pdf", width = 4, height = 4)
display_venn(list, fill = c("red", "blue"))
dev.off()
## pdf
##
```

# Figure 6H

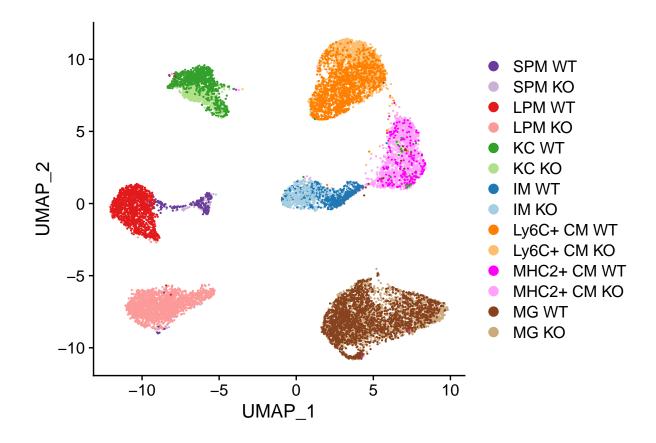
```
MafB_Mac_genes <- intersect(MafB2_genes, Mac_sign_genes)
write.table(append(MafB_Mac_genes, "#MafB_Mac_genes", 0), file = "MafB_Mac_genes.grp",
    row.names = FALSE, col.names = FALSE, quote = FALSE)

path = "C:/Users/domie/Documents/CUTandRUN/GSEA_MafB_Mac_genes"
replotGSEA(path = path, gene.set = "MafB_Mac_genes.grp", class.name = "")
dev.copy(pdf, "C:/Users/domie/Documents/CUTandRUN/GSEA_MafB_Mac_genes.pdf",
    width = 3, height = 3.3)

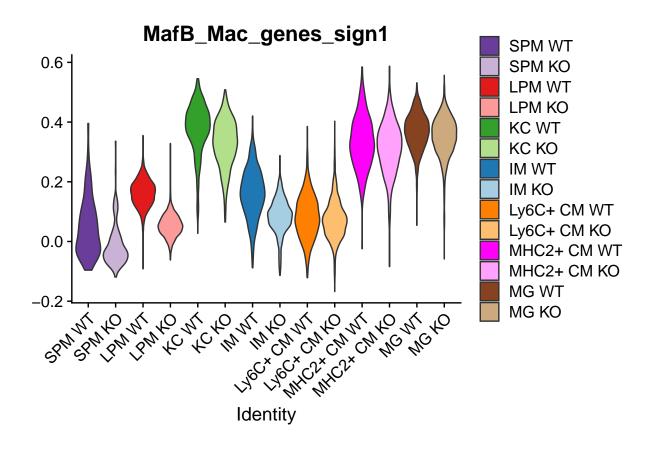
### pdf
##    4

dev.off()

## pdf
##    2</pre>
```

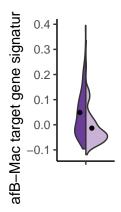


```
sc <- AddModuleScore(sc, features = list(MafB_Mac_genes), name = "MafB_Mac_genes_sign")
VlnPlot(sc, features = "MafB_Mac_genes_sign1", cols = pal, pt.size = 0)</pre>
```



```
metadata <- sc@meta.data
SPM_metadata <- metadata[metadata$cell.type == "SPM", ]
SPM_MafB <- SPM_metadata[, c("Condition", "MafB_Mac_genes_sign1")]

pal_SPM <- c("#6A3D9A", "#CAB2D6")
ggplot(SPM_MafB, aes(x = "", y = MafB_Mac_genes_sign1, fill = Condition)) +
   introdataviz::geom_split_violin() + stat_summary(fun = mean,
        geom = "point", position = position_dodge(width = 0.5), size = 1.5,
        shape = 16, color = "black") + scale_fill_manual(values = pal_SPM) +
        scale_x_discrete(name = "") + ylab("MafB-Mac_target_gene_signature_score") +
        theme classic() + theme(legend.position = "none")</pre>
```



```
ggsave("VlnPlot_Scoring_MafB-Mac_target_SPM.pdf", width = 1.25,
    height = 2)

SPM_WT <- metadata[metadata$Condition == "SPM WT", ]
SPM_KO <- metadata[metadata$Condition == "SPM KO", ]
wilcox.test(SPM_WT$MafB_Mac_genes_sign1, SPM_KO$MafB_Mac_genes_sign1)

##
## Wilcoxon rank sum test with continuity correction
##
## data: SPM_WT$MafB_Mac_genes_sign1 and SPM_KO$MafB_Mac_genes_sign1
## data: SPM_WT$MafB_Mac_genes_sign1 and SPM_KO$MafB_Mac_genes_sign1
## alternative hypothesis: true location shift is not equal to 0</pre>
```

#### LPM

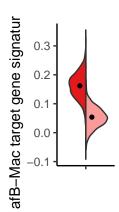
```
metadata <- sc@meta.data
LPM_metadata <- metadata[metadata$cell.type == "LPM", ]
LPM_MafB <- LPM_metadata[, c("Condition", "MafB_Mac_genes_sign1")]

pal_LPM <- c("#E31A1C", "#FB9A99")
ggplot(LPM_MafB, aes(x = "", y = MafB_Mac_genes_sign1, fill = Condition)) +
   introdataviz::geom_split_violin() + stat_summary(fun = mean,</pre>
```

geom = "point", position = position\_dodge(width = 0.5), size = 1.5,
shape = 16, color = "black") + scale fill manual(values = pal LPM) +

theme\_classic() + theme(legend.position = "none")

scale\_x\_discrete(name = "") + ylab("MafB-Mac target gene signature score") +



```
ggsave("VlnPlot_Scoring_MafB-Mac_target_LPM.pdf", width = 1.25,
height = 2)
```

```
LPM_WT <- metadata[metadata$Condition == "LPM WT", ]
LPM_KO <- metadata[metadata$Condition == "LPM KO", ]
wilcox.test(LPM_WT$MafB_Mac_genes_sign1, LPM_KO$MafB_Mac_genes_sign1)</pre>
```

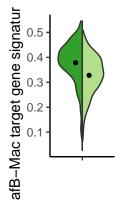
##

```
## Wilcoxon rank sum test with continuity correction
##
## data: LPM_WT$MafB_Mac_genes_sign1 and LPM_KO$MafB_Mac_genes_sign1
## W = 2915366, p-value < 2.2e-16
## alternative hypothesis: true location shift is not equal to 0</pre>
```

#### KC

```
metadata <- sc@meta.data
KC_metadata <- metadata[metadata$cell.type == "KC", ]
KC_MafB <- KC_metadata[, c("Condition", "MafB_Mac_genes_sign1")]</pre>
```

```
pal_KC <- c("#33A02C", "#B2DF8A")
ggplot(KC_MafB, aes(x = "", y = MafB_Mac_genes_sign1, fill = Condition)) +
   introdataviz::geom_split_violin() + stat_summary(fun = mean,
        geom = "point", position = position_dodge(width = 0.5), size = 1.5,
        shape = 16, color = "black") + scale_fill_manual(values = pal_KC) +
        scale_x_discrete(name = "") + ylab("MafB-Mac target gene signature score") +
        theme_classic() + theme(legend.position = "none")</pre>
```



```
ggsave("VlnPlot_Scoring_MafB-Mac_target_KC.pdf", width = 1.25,
height = 2)
```

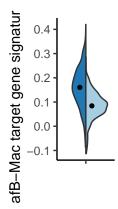
```
KC_WT <- metadata[metadata$Condition == "KC WT", ]
KC_KO <- metadata[metadata$Condition == "KC KO", ]
wilcox.test(KC_WT$MafB_Mac_genes_sign1, KC_KO$MafB_Mac_genes_sign1)</pre>
```

```
##
## Wilcoxon rank sum test with continuity correction
##
## data: KC_WT$MafB_Mac_genes_sign1 and KC_KO$MafB_Mac_genes_sign1
## W = 440476, p-value < 2.2e-16
## alternative hypothesis: true location shift is not equal to 0</pre>
```

IM

```
metadata <- sc@meta.data
IM_metadata <- metadata[metadata$cell.type == "IM", ]
IM_MafB <- IM_metadata[, c("Condition", "MafB_Mac_genes_sign1")]

pal_IM <- c("#1F78B4", "#A6CEE3")
ggplot(IM_MafB, aes(x = "", y = MafB_Mac_genes_sign1, fill = Condition)) +
   introdataviz::geom_split_violin() + stat_summary(fun = mean,
        geom = "point", position = position_dodge(width = 0.5), size = 1.5,
        shape = 16, color = "black") + scale_fill_manual(values = pal_IM) +
        scale_x_discrete(name = "") + ylab("MafB-Mac_target_gene_signature_score") +</pre>
```



```
ggsave("VlnPlot_Scoring_MafB-Mac_target_IM.pdf", width = 1.25,
height = 2)
```

```
IM_WT <- metadata[metadata$Condition == "IM WT", ]
IM_KO <- metadata[metadata$Condition == "IM KO", ]
wilcox.test(IM_WT$MafB_Mac_genes_sign1, IM_KO$MafB_Mac_genes_sign1)</pre>
```

```
##
## Wilcoxon rank sum test with continuity correction
##
## data: IM_WT$MafB_Mac_genes_sign1 and IM_KO$MafB_Mac_genes_sign1
## W = 342906, p-value < 2.2e-16
## alternative hypothesis: true location shift is not equal to 0</pre>
```

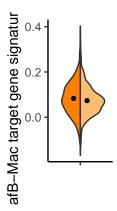
theme\_classic() + theme(legend.position = "none")

#### Ly6C CM

```
metadata <- sc@meta.data
Ly6C_metadata <- metadata[metadata$cell.type == "Ly6C+ CM", ]
Ly6C_MafB <- Ly6C_metadata[, c("Condition", "MafB_Mac_genes_sign1")]

pal_Ly6C <- c("#FF7F00", "#FDBF6F")
ggplot(Ly6C_MafB, aes(x = "", y = MafB_Mac_genes_sign1, fill = Condition)) +
   introdataviz::geom_split_violin() + stat_summary(fun = mean,
   geom = "point", position = position_dodge(width = 0.5), size = 1.5,</pre>
```

```
shape = 16, color = "black") + scale_fill_manual(values = pal_Ly6C) +
scale_x_discrete(name = "") + ylab("MafB-Mac target gene signature score") +
theme_classic() + theme(legend.position = "none")
```



```
ggsave("VlnPlot_Scoring_MafB-Mac_target_Ly6C_CM.pdf", width = 1.25,
height = 2)
```

```
Ly6C_WT <- metadata[metadata$Condition == "Ly6C+ CM WT", ]
Ly6C_KO <- metadata[metadata$Condition == "Ly6C+ CM KO", ]
wilcox.test(Ly6C_WT$MafB_Mac_genes_sign1, Ly6C_KO$MafB_Mac_genes_sign1)
```

```
##
## Wilcoxon rank sum test with continuity correction
##
## data: Ly6C_WT$MafB_Mac_genes_sign1 and Ly6C_KO$MafB_Mac_genes_sign1
## W = 1379681, p-value = 8.946e-06
## alternative hypothesis: true location shift is not equal to 0
```

#### MHC2 CM

```
metadata <- sc@meta.data
MHC2_metadata <- metadata[metadata$cell.type == "MHC2+ CM", ]
MHC2_MafB <- MHC2_metadata[, c("Condition", "MafB_Mac_genes_sign1")]</pre>
```

```
pal_MHC2 <- c("#FF00FA", "#FFA1FD")
ggplot(MHC2_MafB, aes(x = "", y = MafB_Mac_genes_sign1, fill = Condition)) +
  introdataviz::geom_split_violin() + stat_summary(fun = mean,
  geom = "point", position = position_dodge(width = 0.5), size = 1.5,
  shape = 16, color = "black") + scale_fill_manual(values = pal_MHC2) +
  scale_x_discrete(name = "") + ylab("MafB-Mac target gene signature score") +
  theme_classic() + theme(legend.position = "none")</pre>
```

```
afB–Mac target gene signatur
```

```
ggsave("VlnPlot_Scoring_MafB-Mac_target_MHC2_CM.pdf", width = 1.25,
    height = 2)

MHC2_WT <- metadata[metadata$Condition == "MHC2+ CM WT", ]
MHC2_KO <- metadata[metadata$Condition == "MHC2+ CM KO", ]
wilcox.test(MHC2_WT$MafB_Mac_genes_sign1, MHC2_KO$MafB_Mac_genes_sign1)

##
## Wilcoxon rank sum test with continuity correction
##
## data: MHC2_WT$MafB_Mac_genes_sign1 and MHC2_KO$MafB_Mac_genes_sign1
## W = 161609, p-value = 0.01255
## alternative hypothesis: true location shift is not equal to 0</pre>
```

#### MG

```
metadata <- sc@meta.data
MG_metadata <- metadata[metadata$cell.type == "MG", ]
MG_MafB <- MG_metadata[, c("Condition", "MafB_Mac_genes_sign1")]

pal_MG <- c("#87421F", "#CDAA7D")
ggplot(MG_MafB, aes(x = "", y = MafB_Mac_genes_sign1, fill = Condition)) +
   introdataviz::geom_split_violin() + stat_summary(fun = mean,
   geom = "point", position = position_dodge(width = 0.5), size = 1.5,
   shape = 16, color = "black") + scale_fill_manual(values = pal_MG) +
   scale_x_discrete(name = "") + ylab("MafB-Mac target gene signature score") +
   theme_classic() + theme(legend.position = "none")</pre>
```

```
afB–Mac target gene signatur
```

```
ggsave("VlnPlot_Scoring_MafB-Mac_target_MG.pdf", width = 1.25,
    height = 2)

MG_WT <- metadata[metadata$Condition == "MG WT", ]
MG_KO <- metadata[metadata$Condition == "MG KO", ]
wilcox.test(MG_WT$MafB_Mac_genes_sign1, MG_KO$MafB_Mac_genes_sign1)

##
## Wilcoxon rank sum test with continuity correction
##
## data: MG_WT$MafB_Mac_genes_sign1 and MG_KO$MafB_Mac_genes_sign1
## W = 2338163, p-value = 2.299e-15
## alternative hypothesis: true location shift is not equal to 0</pre>
```