Discovery of directional chromatin-associated regulatory motifs

affecting human gene transcription

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Abstract

Abstract

Chromatin interactions are essential in enhancer-promoter interactions (EPIs) and

transcription regulation. CTCF and cohesin proteins located at chromatin interaction

anchors and other DNA-binding proteins such as YY1, ZNF143, and SMARCA4 are

involved in chromatin interactions. However, overall understanding about proteins

associated with chromatin interactions and insulator functions remains unclear. Here, I

describe a systematic and comprehensive approach for discovering DNA-binding motifs

of transcription factors (TFs) that affect EPIs and gene expression. This analysis

identified 96 biased orientations (64 forward-reverse (FR) and 52 reverse-forward (RF))

of motifs that significantly affected the expression level of putative transcriptional target

genes in monocytes, T cells, HMEC, and NPC and included CTCF, cohesin (RAD21

and SMC3), YY1, and ZNF143 (some TFs had more than one motif in databases; hence,

the total number is smaller than the sum of FR and RF). KLF4, ERG, RFX, RFX2,

HIF1, SP1, STAT3, and AP1 were associated with chromatin interactions. Many other

TFs were also known to have chromatin-associated functions. The predicted biased

orientations of motifs were compared with chromatin interaction data. Correlations in

expression level of nearby genes separated by the motif sites were then examined

among 53 tissues. Most TFs showed weak directional biases at chromatin interaction

anchors and were difficult to identify based on enrichment analysis of motifs. These

findings contribute to the investigation of chromatin-associated motifs involved in

transcriptional regulation, chromatin interactions/regulation, and histone modifications.

Keywords: gene expression, transcriptional regulation, transcription factors, enhancer-promoter interactions, chromatin interactions and regulation, CTCF, open chromatin regions, histone modifications

Background

Chromatin interactions play essential roles in enhancer-promoter interactions (EPIs) and transcriptional regulation. CTCF and cohesin proteins are located at the chromatin interaction anchors and facilitate the formation of the loop structures. CTCF functions as an insulator by limiting the activity of enhancers in the loops (Fig. 1a) [1, 2]. The DNA-binding sequence of CTCF indicates its orientation bias at chromatin interaction anchors as the forward-reverse (FR) orientation is frequently observed [1, 3, 4]. Other DNA-binding proteins such as ZNF143, YY1, and SMARCA4 (BRG1) are involved in chromatin interactions and EPIs [5-7]. ZNF143 is known to bind more to the promoters of genes than the distal regions, thereby establishing looping with the distal element bound by CTCF [5]. The DNA-binding motif of ZNF143 is significantly enriched at both chromatin interaction anchors (Z-score > 7), whereas the DNA-binding motif of YY1 is not enriched (Z-score < 2; [5] Figure 2a). CTCF, cohesin, ZNF143, YY1, and SMARCA4 are involved in various biological functions other than chromatin interaction. These transcription factors (TFs) are bound in open chromatin regions near transcriptional start sites (TSS) as well, which may reduce the enrichment of the motifs at chromatin interaction anchors.

Computational approaches to predict chromatin interactions have been developed

[8, 9]. However, they were not intended to find DNA motifs of TFs involved in

chromatin interactions. Xie et al. [10] have concluded that the next challenge will be to

develop systematic methods to discern the specific functions of these motifs in a

genome-wide fashion.

DNA-binding proteins involved in chromatin interactions are considered to affect

the transcription of genes within the loops formed by chromatin interactions. To

confirm this hypothesis, I previously found that the expression level of human putative

transcriptional target genes of a TF was significantly changed according to the

enhancer-promoter association (EPA) [11]. An EPA was shortened at the FR orientation

of CTCF-binding sites, and the transcriptional target genes of each TF bound to

enhancers and promoters were predicted based on the EPA (Fig. 1b) [12]. The

expression levels predicted based on EPA were significantly different compared to the

expression levels predicted based on only promoters (or closed chromatin regions). The

expression levels tended to increase in monocytes and CD4$^+$ T cells, implying that TFs

serve as activators of target genes. The expression levels tended to decrease in

embryonic stem (ES) and induced pluripotent stem (iPS) cells, implying that TFs serve

as repressors of target genes. The expression level of putative transcriptional target

genes of a TF bound in enhancers is changed significantly when EPIs are predicted

properly using the insulator sites of CTCF in different cell types. This conclusion

suggests that the difference in the expression level reflects the accuracy of prediction of

transcriptional target genes based on a given EPA. Using this approach, I planned to

search for other DNA-binding proteins involved in chromatin interactions, which affect

the expression level of putative transcriptional target genes through their insulator

function.

Experimental analyses of chromatin interactions in molecular biology have

technical limitations. Chromatin interaction data are altered according to experimental

techniques, the number of DNA sequencing reads, and even replication sets of the same

cell type. Additionally, chromatin interaction data may not be saturated enough to

contain all chromatin interactions and may include experimental noise such as ligation

errors. To avoid the technical limitations of experimental analyses, a computational

search was conducted for DNA motifs of TFs that affect EPI and the expression level of

putative transcriptional target genes in monocytes, T cells, human mammary epithelial

cells (HMEC), and neural progenitor cells (NPC) without using chromatin interaction

data. Then, I compared putative EPIs and transcription factor binding sites (TFBSs)

with chromatin interaction and gene expression data.

Results

Search for biased orientations of DNA motifs

TFBSs were predicted using open chromatin regions and DNA motifs of TFs

collected from various databases and journals (see Methods). Transcriptional target

genes were predicted based on TFBSs in promoters and enhancers. An EPA was

shortened at the TFBSs with insulator function, such as CTCF and cohesin (RAD21 and

SMC3). To find motifs of TFs with insulator function other than CTCF and cohesin, an

EPA was shortened at the TFBSs of a putative insulator TF. Transcriptional target genes

of each general TF bound in (i) distal open chromatin regions (enhancers) and (ii)

closed chromatin regions in promoters were predicted based on the EPA in order to

estimate the enhancer activity of each general TF. The ratio of expression level between

(i) and (ii) was calculated. The distribution of the ratios of all general TFs was

compared among forward-reverse (FR), reverse-forward (RF) orientations, and

non-orientation (i.e., without considering orientation) groups of TFBSs using a

two-sided Mann-Whitney test (p-value $< 10^{-7}$). Putative insulator TFs with significant

differences in the distributions of the ratios of expression level among FR, RF, and

non-orientation were enumerated (Fig. 2a).

To confirm the predictions of TFBSs, the distributions of TFBSs were examined

in the human genome. TFBSs involved in chromatin interactions were identified near

TSS (Fig. 2b). ZNF143 is known to bind to the promoter of genes, thereby establishing

looping with the distal element bound by CTCF [5]. YY1 is a multi-functional

transcriptional regulator and is involved in transcriptional activation, repression, and

initiation [13, 14]. The distributions of TFBSs seemed to be consistent with the features

of the TFs. Most TFBSs were identified in intergenic and intron regions of the human

genome (Fig. 2c).

Several hundreds of FR and RF orientations were found in monocytes, T cells,

HMEC, and NPC (Fig. 3). The number of the TFs seemed to be changed, according to

the number of DNase-seq reads, since the number of the reads in monocytes was larger

than those of the other three cell types. When DNase-seq reads increase, more TFBSs

may be predicted to some extent. Ninety-six biased (sixty-four FR and fifty-two RF)

orientations of DNA-binding motifs of TFs were discovered and found to be common

between the four cell types, whereas non-orientation of DNA-binding motifs of TFs

were not detected (Fig. 3; Table 1; Additional file 1: Table S1). The FR orientation of

DNA motifs included CTCF, cohesin (RAD21 and SMC3), YY1, and ZNF143, which

are involved in chromatin interactions and EPIs.

Without considering the difference of orientations among cell types, 175 biased

orientations were found in common between the four cell types. These numbers (96 and

175) are the numbers of unique TFs with biased orientations of DNA motifs and not the

numbers of unique DNA motifs. Possible causes of the discrepancy are as follows: (1)

A TF or an isoform of the same TF may bind to a different DNA-binding motif,

according to cell types and/or in the same cell type. About 50% or more alternative

splicing isoforms are differently expressed among tissues, thereby indicating that most

alternative splicing is subject to tissue-specific regulation [15-17]; (2) The same TF has

several DNA-binding motifs, and in some cases, one of the motifs is almost the same as

the reverse complement motif of another motif of the same TF in a database; (3) I

previously found that a complex of TFs would bind to a slightly different DNA-binding

motif by combining DNA-binding motifs from the complex in *C. elegans* [18]. In other

reports, the difference of binding specificity of TF pairs was investigated [19].

Cofactors also alter the binding specificity of TFs [20-22]. These increase the variety

and number of (biased orientations of) DNA-binding motifs, which are potentially

involved in different biological functions among cell types.

Among the 96 TFs, KLF4 [23], ERG [24, 25], RFX [26], RFX2 [27], HIF1 [28],

SP1 [29], STAT3 [30], and AP1 [31] were found to be associated with chromatin

interactions. Myc [32, 33], the ETS family, ELK3 [34], ELK4 [35], RUNX3 [36], USF1

[37, 38], AHR/ARNT [39], SP2 [40], EGR2 [41-43], ZBTB33 [44], NF-Y [45, 46],

MITF [47, 48], NONO [49], FLI1 [50], E2F [51-53], BCL [54], CXXC1 [55], E2F1

[56], EGR1 [57], REST [58, 59], RFX [60], SIN3A [61-63], AP1 [64], and STAT3 and

STAT5 [65] were associated with chromatin remodeling, chromatin accessibility, and/or

histone modification. NRF1 [6, 66] and TFAP2 [67] were pioneer factors, which were

associated with chromatin remodeling and accessibility. TFBSs of the TFs other than

those involved in chromatin interactions would affect gene expression by

chromatin-associated functions. Huang et al. [37] have mentioned that the insulator

element at the 5' end of the chicken beta-globin locus acts as a barrier, thus protecting

transgenes against silencing effects of adjacent heterochromatin. The transcription

factor USF1 binds within the insulator at a site important in adjacent nucleosomes for

generating histone modifications associated with active chromatin and, by inference,

with barrier function.

Comparison with chromatin interaction data

To examine whether the biased orientations of DNA motifs were associated with

chromatin interactions, EPIs predicted based on an EPA were compared with chromatin

interaction data. EPIs predicted based on an EPA were compared with three replications

(B2T1, B2T2, and B3T1) of HiChIP chromatin interaction data in CD4$^+$ T cells

separately [68]. The resolutions of HiChIP chromatin interaction data and EPIs were

adjusted to 5 kilobases (kb). The number of unique HiChIP cis (i.e., in the same

chromatin) chromatin interactions was 19,926,360 at 5-kb resolution in B2T1

replication of T cells; 666,149 at 5-kb resolution with $\geq 1,000$ counts for each

interaction; and 78,209 at 5-kb resolution with $\geq 6,000$ counts for each interaction

(Additional file 1: Table S2). EPIs were predicted based on three types of EPAs: (i)

EPA shortened at the FR or RF orientation of TFBSs of a TF with putative insulator

function such as CTCF, (ii) EPA shortened at the TFBSs of a TF without considering

their orientation, and (iii) EPA which were not shortened by TFBSs. For a total of 201

biased orientations [133 of 273 (49%) FR and 90 of 211 (43%) RF] of DNA motifs in T

cells, which included CTCF, cohesin (RAD21 and SMC3), ZNF143, and YY1 in three

replications (B2T1, B2T2, and B3T1), a significantly higher ratio of EPIs overlapped

with HiChIP chromatin interactions with $\geq 1,000$ counts for each interaction in EPA (i)

than in EPA (ii) and (iii) in T cells (Table 2). For a total of 390 biased orientations [261

of 273 (96%) FR and 199 of 211 (94%) RF] of DNA motifs in T cells, a significantly

higher ratio of EPIs overlapped with the chromatin interactions in EPA (i) than EPA

(iii) (Table 2; Additional file 1: Table S3). The difference between EPIs predicted based

on EPA (i) and (ii) seemed to be difficult to distinguish using the chromatin interaction

data and a statistical test in some cases. However, for the difference between EPIs

predicted based on EPA (i) and (iii), a larger number of biased orientations of DNA

motifs were correlated with chromatin interaction data. Most of the biased orientations

of DNA motifs (95%) were correlated with chromatin interactions when comparing

EPIs predicted based on EPA (i) and (iii) with HiChIP chromatin interactions.

Chromatin interaction data were obtained from different samples of DNase-seq open

chromatin regions; thus, individual differences may exist. Individual differences of

chromatin interactions were larger than those of open chromatin regions [68].

Additionally, to confirm these tendencies of comparison of EPIs with HiChIP

chromatin interactions, the same analysis was conducted using HiChIP chromatin

interaction data in Th17, Treg, and GM12878 cells. For a total of 128 biased

orientations [76 of 200 (38%) FR and 64 of 183 (35%) RF] of DNA motifs in Th17

cells, which included CTCF, cohesin (RAD21 and SMC3), ZNF143, and YY1 in three

replications (B1T2, B2T1, and B3T1), a significantly higher ratio of EPIs overlapped

with HiChIP chromatin interactions in EPA (i) than in EPA (ii) and (iii) (Additional file

1: Table S4). When comparing EPIs predicted based only on EPA (i) and (iii) with the

chromatin interactions, for a total of 282 biased orientations [182 of 200 (91%) FR and

162 of 183 (89%) RF] of DNA motifs, a significantly higher ratio of EPIs overlapped

with the chromatin interactions in EPA (i) than EPA (iii) (Additional file 1: Table S4).

In Treg cells, chromatin interactions in one biological replication were overlapped

with EPIs, but because those in the other two biological replications were less

overlapped with EPIs, one replication (B3T1) was used for this analysis. For a total of

281 biased orientations [154 of 290 (53%) FR and 176 of 323 (54%) RF] of DNA motif

sequences, which included CTCF, cohesin (RAD21 and SMC3), ZNF143, and YY1, a

significantly higher ratio of EPIs overlapped with HiChIP chromatin interactions in

EPA (i) than the other types of EPA (ii) and (iii) (Additional file 1: Table S4). When

comparing EPIs predicted based on only EPA (i) and (iii) with the chromatin

interactions, for total 482 biased orientations [283 of 290 (98%) FR and 309 of 323

(96%) RF] of DNA motif sequences, a significantly higher ratio of EPIs overlapped

with the chromatin interactions in EPA (i) than EPA (iii) (Additional file 1: Table S4).

In GM12878 cells, for a total of 168 biased orientations [96 of 229 (38%) FR and

98 of 237 (35%) RF] of DNA motifs including CTCF, cohesin (RAD21 and SMC3),

ZNF143, and YY1 in two replications (B1 and B2), a significantly higher ratio of EPIs

overlapped with HiChIP chromatin interactions in EPA (i) than EPA (ii) and (iii)

(Additional file 1: Table S4). When comparing EPIs predicted based on only EPA (i)

and (iii) with the chromatin interactions, for a total of 366 biased orientations [218 of

229 (95%) FR and 227 of 237 (96%) RF] of DNA motifs, a significantly higher ratio of

EPIs overlapped with the chromatin interactions in EPA (i) than EPA (iii) (Additional

file 1: Table S4).

In situ Hi-C data in HMEC, which are available in a 4D nucleome data portal,

were compared with putative EPIs. The number of unique in situ Hi-C cis chromatin

interactions was 121,873,642 at 3-kb resolution. For a total of 222 biased orientations

[155 of 310 (50%) FR and 108 of 269 (40%) RF] of DNA motifs including CTCF,

cohesin (RAD21 and SMC3), ZNF143, and YY1, a significantly higher ratio of EPIs

overlapped with in situ Hi-C chromatin interactions ($\geq$1 count for each interaction, i.e.,

all interactions) in EPA (i) than in EPA (ii) and (iii) (Additional file 1: Table S5). When

comparing EPIs predicted based on only EPA (i) and (iii) with the chromatin

interactions, for a total of 406 biased orientations [268 of 310 (86%) FR and 230 of 269

(86%) RF] of DNA motifs, a significantly higher ratio of EPIs overlapped with the

chromatin interactions in EPA (i) than EPA (iii) (Additional file 1: Table S5). There

were six replications of in situ Hi-C experimental data of HMEC, but other replications

contained much smaller number of chromatin interactions. The HiChIP chromatin

interaction data in T cells were expected to be enriched with EPIs, since they were

produced with H3K27ac antibody to capture chromatin interactions. In situ Hi-C data

consist of all types of chromatin interactions and thus contain a smaller number of EPIs

than HiChIP data. Chromatin interactions with a small number of counts may be

artifacts caused by ligation errors during experiments [69, 70]. Therefore, other

replications of in situ Hi-C data with a smaller number of chromatin interactions will

not be available for this analysis.

Promoter capture Hi-C data of NPC were downloaded and analyzed by the same

procedure as that for T cells and HMEC [71]. The number of unique promoter capture

Hi-C cis chromatin interactions was 9,436,689 at 5-kb resolution. For a total of 109

biased orientations [61 of 205 (28%) FR and 60 of 241 (25%) RF] of DNA motifs,

which included CTCF, cohesin (RAD21 and SMC3), ZNF143, and YY1, a significantly

higher ratio of EPIs overlapped with promoter capture Hi-C chromatin interactions (≥ 1

count for each interaction, i.e. all interactions) in EPA (i) than in EPA (ii) and (iii)

(Additional file 1: Table S6). When comparing EPIs predicted based on only EPA (i)

and (iii) with the chromatin interactions, for a total of 186 biased orientations [99 of 205

(48%) FR and 103 of 241 (43%) RF] of DNA motifs, a significantly higher ratio of

EPIs overlapped with the chromatin interactions in EPA (i) than EPA (iii) (Additional

file 1: Table S6).

All three types of chromatin interaction data (HiChIP, in situ Hi-C, and promoter

capture Hi-C) indicated that EPIs predicted based on EPA (i) that was shortened at

TFBSs of known TFs involved in chromatin interactions (CTCF, RAD21, SMC3,

ZNF143 and YY1) overlapped with more chromatin interactions than those based on

EPA (ii) and (iii) in six cell types (T cells, Th17, Treg, GM12878, HMEC, and NPC). In

monocytes, Hi-C data was available, but it was not analyzed due to the low resolution of

chromatin interaction data (50 kb).

The biased orientations of TFBSs of known TFs involved in chromatin

interactions such as CTCF, RAD21, SMC3, YY1, and ZNF143 might overlap with the

biased orientations of TFBSs of other TFs; thus, they showed a significantly higher ratio

of EPIs overlapped with chromatin interactions in EPA (i) than in EPA (ii) and (iii).

However, among 273 FR and 211 RF orientations of DNA motifs of TFs not including

the known TFs in T cells, the TFBSs of only 15 (6%) FR and 9 (4%) RF orientations of

DNA motifs of TFs overlapped with $\geq$ 60% of the TFBSs of biased orientations of

DNA motifs of either of the known TFs at 5-kb resolution.

Moreover, to investigate the enhancer activity of EPIs, the distribution of

expression level of putative target genes of EPIs was compared between EPIs

overlapped with HiChIP chromatin interactions and EPIs not overlapped with them.

Though the target genes of EPIs were selected from top 4,000 transcripts (genes) in

terms of the expression level of all transcripts (genes) excluding transcripts not

expressed in T, Th17, Treg, and GM12878 cells separately, target genes of EPIs

overlapped with chromatin interactions indicated a significantly higher expression level

than non-overlapped EPIs, suggesting that EPIs overlapped with chromatin interactions

activated the expression of target genes in the four cell types. For almost all

(99%-100%) FR and RF orientations of DNA motifs, a significantly higher expression

level of putative target genes was observed for EPIs overlapped with chromatin

interactions than for EPIs not overlapped in the four cell types. When a biased

orientation of DNA motifs showed a significantly higher expression level of putative

target genes of EPIs overlapped with chromatin interactions than EPIs not overlapped,

'1' was marked with in the tables of comparison between EPIs and HiChIP chromatin

interactions (Additional file 1: Table S3). DNA motifs showing a significantly lower

expression level or not showing a significant difference of expression level were not

observed in this analysis. HiChIP data were produced using H3K27ac antibody; thus,

chromatin interactions acting as repressors would not be identified. However, for in situ

Hi-C and promoter capture Hi-C data in HMEC and NPC, the significant difference of

expression level of putative target genes of EPIs was observed in a small percentage

(less than 1%) of biased orientations of DNA motifs.

Recently, high resolution chromatin interaction data became publicly available

[72]. The biased orientations of TFBSs of TFs were examined using Micro-C chromatin

interaction data in human fibroblasts. The resolution of the data was adjusted to 200 bp.

TFBSs of CTCF, RAD21, and SMC3 showed a significant directional bias at the pairs

of chromatin interaction anchors. SMRC2-binding sites showed a directional bias at the

anchors (p-value $< 10^{-50}$). SMARCA4 is known to be involved in higher-order

chromatin structures such as the topologically associating domain (TAD) [7]. DNA

motifs of SMARCA4 did not exist in the databases of DNA motifs of TFs used in this

analysis. SMRC2 was not identified in this analysis because TFBSs of TFs affecting

EPI and gene expression were searched, which are not high-order chromatin structures.

ATF3 and ZBTB7A in Table 1 were found among the top 32 biased orientations of

TFBSs of TFs at the pairs of chromatin interaction anchors with p-value $< 10^{-50}$

(Additional file 1: Table S8). Deletion of an ATF3-binding site affected the expression

of a target gene, which was far apart from the site [73]. ZBTB7A is involved in the

changes in chromatin accessibility and the regulation of gene expression [74]. TFBSs of

YY1 and ZNF143 also indicated a weak directional bias at the chromatin interaction

anchors (Additional file 1: Table S7). In this analysis, biased orientations of TFBSs of

TFs were investigated based on the difference of distribution of the expression level of

putative transcriptional target genes. YY1 and ZNF143 were found to have significant

directional biases, thereby implying that the analysis in this study could find TFs that

did not have a strong directional bias based on the number of TFBSs at chromatin

interaction anchors. The other TFs, except for CTCF, RAD21, SMC3, ATF3, and

ZBTB7A in Table 1 showed weak directional biases at the chromatin interaction

anchors (Additional file 1: Table S7).

Correlation of expression level of gene pairs separated by TFBS

To investigate whether the biased orientations of TFBSs affected gene expression,

the expression level of protein-coding genes in monocytes, that are most closely located

upstream and downstream of any biased orientations of TFBSs in the human genome,

was compared. The expression levels of closely located genes including divergent gene

pairs tend to be correlated among tissues [75, 76]. When a biased orientation of TFBSs

serves as an insulator (e.g. CTCF), the correlation of gene expression level would be

reduced. Among 96 biased (64 FR and 52 RF) orientations of TFBSs found in common

in monocytes, T cells, HMEC, and NPC, TFs with ≥50 genomic locations of the

TFBSs were selected after the elimination of ubiquitously expressed genes among 53

tissues and TFBSs near the genes with a coefficient of variance (CV) < 90. After this

filtering, 43 of 60 (72%) FR and 40 of 50 (80%) RF orientations showed significantly

lower correlations of expression levels of gene pairs separated by the TFBSs of a TF

than the correlations of all pairs of neighboring genes with intergene distance <1

megabase (Mb) (Mann-Whitney test, *p*-value < 0.05) (Table 3). The FR and RF

orientations of TFs included known TFs involved in chromatin interactions such as

CTCF, RAD21, SMC3, YY1, and ZNF143. The top 20 FR and RF orientations of DNA

motifs were selected in ascending order of the median correlation coefficient of

expression levels of gene pairs separated by the TFBSs of a TF. The 20 TFs included

known TFs involved in chromatin interactions such as CTCF, RAD21, and SMC3 (Fig.

4) (Mann-Whitney test, *p*-value < 0.01). Instead of using all TFBSs of a TF, when

TFBSs of a TF were limited to sites used in the analysis of biased orientations of DNA

motifs in an EPA, 33 of 55 (60%) FR and 36 of 47 (77%) RF orientations showed

significantly lower correlations of expression levels of gene pairs separated by the

TFBSs (Additional file 1: Table S8). With intergene distance < 20 kb, 21 of 62 (34%)

FR and 8 of 52 (15%) RF orientations showed significantly lower correlations of

expression levels of gene pairs separated by the TFBSs of a TF than the correlations of

all pairs of neighboring genes (Additional file 1: Table S8).

A large number of TFBSs of a TF tend to show significantly lower correlations of

gene expression (median numbers of TFBSs showing significantly lower correlations

were 656 and 479 for FR and RF respectively, and median numbers of TFBSs not

showing significantly lower correlations were 122 and 115 for FR and RF respectively.

The Mann-Whitney test p-value $= 0$ and $< 10^{-10}$ for FR and RF respectively). This

suggests that if the criteria to examine the correlations of expression levels of gene pairs

separated by TFBSs are more stringent and if the number of TFBSs is reduced, it would

be more difficult to use a statistical test to compare the correlations of expression levels

of gene pairs separated by the reduced TFBSs, particularly for TFs with a small number

of TFBSs.

Among 91 biased orientations of DNA motifs of TFs found in common among

the four cell types, after excluding the five known TFs involved in chromatin

interactions such as CTCF, RAD21, SMC3, YY1, and ZNF143, TFBSs of 88 TFs were

located at ≥ 1 genomic regions between gene pairs not including TFBSs of the known

TFs and at ≥ 1 genomic regions between low correlation (< 0.13) of expression level of

gene pairs. Ubiquitously expressed genes among 53 tissues with CV < 90 were

eliminated. The median correlation coefficient of expression level of all pairs of nearby

genes was 0.17 (Table 3). TFBSs of 52 TFs were placed at ≥ 10 genomic regions.

Co-location of biased orientations of motifs

To examine the association of biased orientations of DNA-binding motifs of TFs,

co-location of motifs within 200 bp (approximately the length of the minimum open

chromatin region of 146 bp) of each other was analyzed in monocytes, T cells, HMEC,

and NPC. The same pairs of motifs upstream and downstream of genes in EPAs were

enumerated, and the pairs of motifs found in ≥ 100 genomic regions were listed (Table

4; Additional file 1: Table S9). A majority of the pairs of motifs overlapped with more

than 1 base, but the same tendency was found in the analysis of motifs of all TFs

(Additional file 1: Table S9). CTCF partially overlapped with cohesin RAD21 and

SMC3 (Table 4). The top 30 pairs of FR and RF orientations of motifs overlapped or

co-localized have been shown (Table 4). Unique pairs of overlapping or co-localized

motifs were 404 FR and 662 RF, consisting of 117 FR and 218 RF of unique DNA

motifs in monocytes. Only 10 overlapping or co-localized motifs were found to be

common among the four cell types, including pairs of CTCF, RAD21, SMC3, and YY1,

thereby implying that most overlapping and co-localized motifs serve in a cell-specific

manner (Additional file 1: Table S9). Paired DNA-binding motifs of TFs with biased

orientations tend to overlap, and a relatively small number of motifs are co-localized

upstream and downstream of genes. Overlapping motifs act as TFBSs for several TFs

either simultaneously or competitively [20-22, 77-81]. CTCF and cohesin function

cooperatively with another TF and a protein such as Klf4 and Brd2 [82, 83]. Klf4

recruits cohesin to the Oct4 enhancer region. BRD2 supports boundary (insulator)

activity of CTCF. The analysis of 457 ChIP-seq data sets from 119 human TFs showed

that secondary motifs were detected in addition to the canonical motifs of the TFs,

indicating tethered binding and co-binding between multiple TFs [84]. Wang et al. [84]

observed significant position and orientation preferences between many co-binding TFs.

Overlapping motifs found in this study may be biologically functional rather than

artifact. Though repeat sequences in the human genome were masked, more than a

hundred pairs of motifs overlapped with hundreds or thousands of upstream and

downstream gene regions.

Discussion

An initial comprehensive method to find DNA-binding motifs of TFs affecting

EPIs and gene expression without using chromatin interaction data is presented through

this study. DNA motifs, such as CTCF and cohesin (RAD21 and SMC3), were searched

from the open chromatin regions of monocytes, T cells, HMEC, and NPC. A total of 96

biased (64 FR and 52 RF) orientations of DNA motifs of TFs were found to be common

among the four cell types. Non-orientation (i.e., without considering orientation) of

motifs was not found to affect the expression level of putative transcriptional target

genes, suggesting that an EPA shortened at the FR or RF orientation of TFBSs was a

more accurate prediction of EPIs. This was supported by the comparisons with

chromatin interaction and gene expression data. FF and RR orientations of motifs were

not examined in this analysis since CTCF has FR or RF orientation of its TFBSs.

Among 96 biased orientations of DNA motifs of TFs, 14 TFs were reported to be

associated with chromatin interactions, and many other TFs had chromatin-associated

functions. The enrichment analysis of biased orientations of TFBSs of the 96 TFs, for

pairs of chromatin interaction anchors indicated a weak statistical significance, except

for CTCF, RAD21, SMC3, ATB3, and ZBTB7A, suggesting that the TFs were difficult

to find using DNA motifs of TFs [5].

To predict TFBSs in genomic DNA, several tools and methods were explored and

then protein interaction quantitation (PIQ) was used [66]. PIQ predicts TFBSs by

reducing noise in DNase-seq data, and the comparison of the prediction with ChIP-seq

data has revealed that PIQ shows a better performance than other tools and digital

genome footprinting (DNase-DGF). The prediction of TFBSs using PIQ in this study

increased the number of biased orientations of motifs, thereby reducing the number of

non-orientation of motifs. At first, the number of predicted motifs was expected to

increase relative to using consensus DNA-binding sequences of TFs and open

chromatin regions available in ENCODE database, which seemed to be difficult to find

all TFBSs in the genome. However, the number of TFBSs was almost the same or

slightly decreased compared with the analysis of consensus sequences, suggesting that

PIQ increased true-positive predictions of TFBSs and reduced false-positive

predictions.

When forming a homodimer or heterodimer with another TF, TFs may bind to

genomic DNA with a specific orientation of their TFBSs. The approach in this study

would be able to identify TFs forming a heterodimer. If the TFBS of only the mate to a

pair of TFs was found in an EPA, the EPA was shortened at one side of the TFBS, and

transcriptional target genes were predicted based on the EPA shortened at the side. In

this analysis, the mate to both heterodimers and homodimers of TFs can be tested to

examine the effect on the expression level of transcriptional target genes predicted

based on the EPA shortened at one side. Biased orientations of DNA-binding motifs

may also be found in forward-forward or reverse-reverse orientation.

TFs that are multi-functional transcriptional regulators such as YY1 may not

indicate a strong directional bias of their TFBSs. Though some proportion of TFBSs of

a TF involved in chromatin interactions and insulator functions is found in a biased

orientation in the genome, the other TFBSs of the TF may not form a dimer with

another TF and may not be present in a specific orientation in the genome. Thus, the

overall proportion of biased orientations of TFBSs of a TF may be reduced such that a

statistical test cannot detect it.

The approach presented in this study was tested to search for repeat sequences

that serve as insulator sites and affect the expression level of putative transcriptional

target genes of TFs. A small number of repeat sequences were identified and included

MIR, which act as insulators [85].

CTCF and cohesin-binding sites are frequently mutated in cancer [86]. Some

biased orientations of DNA motifs involved in chromatin interactions may be associated

with diseases including cancer.

The analyses conducted here suggest that biased orientations of DNA motifs of

discovered TFs affect EPIs and the expression level of putative transcriptional target

genes of general TFs bound in enhancers and promoters. This gives insight into the

nature of TFs and contributes to the research of functions and molecular mechanisms of

chromatin interactions, regulation of chromatin, histone modifications, transcriptional

regulation, and transcriptional cascades.

Conclusion

CTCF and cohesin proteins are located at the anchors of chromatin interactions

and form loop structures. Other DNA-binding proteins such as YY1, ZNF143, and

SMARCA4 are also reported to be involved in chromatin interactions. However, the

overall understanding of proteins associated with chromatin interactions and insulator

function remains unclear. The analysis performed here discovered a hundred of FR and

RF orientations of DNA motifs of TFs which affected EPIs and gene expression among

four cell types. However, without considering the orientations, DNA motifs of TFs

affecting EPIs and gene expression were not detected. Among the biased orientations of

TFs, KLF4, ERG, RFX, RFX2, HIF1, SP1, STAT3, and AP1 were reported to be

associated with chromatin interactions. Many other TFs were also known to have

chromatin-associated functions. Most of the TFs showed weak directional biases at

chromatin interaction anchors and were difficult to identify based on their enrichment

analysis of TFBSs. These findings contribute to investigate chromatin-associated motifs

involved in transcriptional regulation, chromatin interactions, regulation of chromatin,

and histone modifications.

Methods

Search for biased orientations of DNA motifs

Methods

Search for biased orientations of DNA motifs

To identify transcription factor binding sites (TFBSs) from open and closed

chromatin regions, TRANSFAC (2019.2), JASPAR (2018), UniPROBE (2018),

high-throughput SELEX, transcription factor binding sequences of ENCODE ChIP-seq

data, and HOCOMOCO version 9 and 11 were used to predict insulator sites [87-94].

TRANSFAC (2019.2) was used to analyze enhancer-promoter interactions because

these data were sufficient to identify biased orientations of DNA motifs of insulator TFs

with less computational time, thereby reducing the number of non-orientation of DNA

motifs of TFs. To reduce false-positive predictions of TFBSs in open chromatin regions,

the DNA motif search was performed by using the protein interaction quantitation (PIQ)

tool with 12,249 position frequency matrices converted from DNA motifs of vertebrate

TFs in the above-mentioned databases and DNase-seq data from GEO and ENCODE

database (GSM1024791 CD14$^+$ monocytes; GSM665812 CD4$^+$ T cell; GSM736634

Human mammary epithelial cell, HMEC; GSM878615 Neural progenitor cell, NPC;

ENCFF263GMV, ENCFF886XEV Th17; GSM1024741 Treg; GSM736620 GM12878)

[66]. To find open chromatin regions associated with enhancer and promoter activity,

narrow peaks of ChIP-seq experimental data of H3K27ac were predicted using MACS2

callpeak (GEO: GSM773004) [95]. To predict TFBSs of TFs from closed chromatin

regions, another method was employed. Position weight matrices of vertebrate

transcription factor binding sequences were converted into DNA consensus sequences,

which consist of an alphabet of 15 characters (the four bases A, C, G, T, the six

two-fold degenerate IUPAC codes R=[AG], Y=[CT], K=[GT], M=[AC], S=[GC],

W=[AT], the four three-fold degenerate IUPAC codes B=[CGT], D=[AGT], H=[ACT],

V=[ACG], and the four-fold degenerate character N=[ATGC]) using a convert matrix in

the RSAT website [96]. Transcription factor binding sequences were searched from

narrow peaks of DNase-seq data in repeat-masked hg19 genome sequences using the

Match tool in TRANSFAC with a similarity score cutoff of 1 after transforming the

DNA consensus sequences of TFs into position frequency matrixes for the Match search

[97]. Repeat-masked hg19 genome sequences were downloaded from the UCSC

genome   browser   (http://genome.ucsc.edu/,

http://hgdownload.soe.ucsc.edu/goldenPath/hg19/bigZips/hg19.fa.masked.gz).   TFs

corresponding to transcription factor binding sequences were searched computationally

and manually by comparing their names and gene symbols of HUGO Gene

Nomenclature Committee-approved gene nomenclature and 31,848 UCSC canonical

transcripts

(http://hgdownload.soe.ucsc.edu/goldenPath/hg19/database/knownCanonical.txt.gz)

because transcription factor binding sequences were not linked to transcript IDs such as

UCSC, RefSeq, and Ensembl transcripts.

Target genes of a TF were assigned when its TFBS was found in the DNase-seq

narrow peaks in the promoter or extended regions of a gene's EPA. Promoter and

extended regions were defined as follows: promoter regions were those that were within

a distance of ±5 kb from transcriptional start sites (TSS). Promoter and extended

regions were defined as per the following association rule, which is the same as that

defined in Figure 3A of a previous study [12]: the single nearest gene association rule,

which extends the regulatory domain to the midpoint between the TSS of the gene and

that of the nearest gene upstream and downstream without the limitation of extension

length. The limitation of extension length was ＜1 Mb in this study. Extended regions

for an EPA were shortened at the TFBSs of a putative insulator TF closest to a TSS, and

transcriptional target genes were predicted based on the shortened EPA and TFBSs of a

general TF in open chromatin regions. Promoter and extended regions of an EPA were

shortened at the forward–reverse (FR) orientation of TFBSs of a putative insulator TF.

When the forward or reverse orientation of TFBSs was located in the genome several

times, the most internal (i.e., closest to a TSS) forward–reverse orientation of the

TFBSs was selected. The genomic positions of genes were identified using the

'knownGene.txt.gz' file in UCSC bioinformatics sites [98]. The file

'knownCanonical.txt.gz' was also utilized to select representative transcripts among

various alternate forms for assigning promoter and extended regions of an EPA. From

the list of TFBSs and transcriptional target genes, redundant DNA motifs of TFs were

removed by comparing the putative target genes of a TFBS and its corresponding TF; if

identical, one of the DNA motifs of TF was chosen. When the number of transcriptional

target genes predicted based on a TFBS was <5, the DNA motif of TF was omitted.

For gene expression data, RNA-seq reads mapped onto the human hg19 genome

sequence were obtained from UCSF-UBC human reference epigenome mapping project

RNA-seq reads with poly-A of naive CD4$^+$ T cells (GEO: GSM669617). For monocytes,

Blueprint   RNA-seq   FPKM   data

('C0010KB1.transcript_quantification.rsem_grape2_crg.GRCh38.20150622.results')

were   downloaded   from   Blueprint   DCC   portal

(http://dcc.blueprint-epigenome.eu/#/files). RNA-seq reads with the poly-A of human

mammary epithelial cells (HMEC) in the Encyclopedia of DNA Elements at UCSC

(http://genome.ucsc.edu/encode/,

'wgEncodeCshlLongRnaSeqHmecCellPapAlnRep1.bam' file), H1-derived NPC (GEO:

GSM915326, ENCODE: ENCFF529SIO), Th17 (GSM2859479, NCBI SRA:

SRR6298326), Treg (GSM2859476, SRR6298323), and GM12878 (GSE78551,

ENCFF297QCE) were analyzed. RNA-seq reads were aligned in repeat-masked hg19

genome sequences using Burrows-Wheeler Aligner with default parameters [99].

FPKMs of the RNA-seq data were calculated by RSeQC [100]. Based on

log2-transformed FPKM, UCSC transcripts were arranged in a descending order of

expression level, and the top 4,000 transcripts were selected in each cell type.

The expression level of transcriptional target genes predicted based on EPAs

shortened at the TFBSs of a TF was compared with the expression level of

transcriptional target genes predicted based on closed chromatin regions in promoters.

For each DNA motif shortening EPAs, transcriptional target genes were predicted based

on DNA-binding motifs of vertebrate TFs in the TRANSFAC database, and the ratio of

expression level of putative transcriptional target genes of each TF was calculated

between an EPA and closed chromatin regions in the promoters. The distribution of the

ratios of all TFs was compared among forward-reverse (FR), reverse-forward (RF), and

non-orientation (i.e., without considering orientation) of TFBSs shortening an EPA

using a two-sided Mann-Whitney test, (p-value $< 10^{-7}$). Other parameters for the

analysis were set as follows: the number of transcriptional target genes of a TF was

≥50, the number of general TFs (not insulator TFs) to predict transcriptional target

genes was ≥50, and the number of TFBSs of a TF (insulator TF) shortening an EPA

was ≥100.

Comparison with chromatin interaction data

For comparison of EPIs with chromatin interactions (HiChIP) in CD4$^+$ T, Th17,

Treg, and GM12878 cells,

'GSM2705049_Naive_HiChIP_H3K27ac_B2T1_allValidPairs.txt',

'GSM2705050_Naive_HiChIP_H3K27ac_B2T2_allValidPairs.txt',

'GSM2705051_Naive_HiChIP_H3K27ac_B3T1_allValidPairs.txt',

'GSM2705054_Th17_HiChIP_H3K27ac_B1T2_allValidPairs.txt',

'GSM2705055_Th17_HiChIP_H3K27ac_B2T1_allValidPairs.txt',

'GSM2705056_Th17_HiChIP_H3K27ac_B3T1_allValidPairs.txt',

'GSM2705059_Treg_HiChIP_H3K27ac_B3T1_allValidPairs.txt',

'GSM2705041_GM_HiChIP_H3K27ac_B1_allValidPairs.txt', and

'GSM2705042_GM_HiChIP_H3K27ac_B2_allValidPairs.txt' files were downloaded

from the Gene Expression Omnibus (GEO) database. The resolutions of chromatin

interactions and EPIs were adjusted to 5 kb before comparison. Chromatin interactions

with ≥ 6,000 and ≥ 1,000 counts for each interaction were analyzed in the comparison

in CD4$^+$ T cells. According to the unique numbers of chromatin interactions in

replications in Th17, Treg, and GM12878 cells, chromatin interactions with ≥1,000

(B1T2 in Th17), ≥2,000 (B2T1 and B3T1 in Th17), ≥ 6,000 (B3T1 in Treg), and

≥4,000 (B1 and B2 in GM12878) counts for each interaction were analyzed in the

comparison.

EPIs were predicted on the basis of the following three types of EPAs in

monocytes: (i) EPA shortened at the FR or RF orientation of TFBSs of a TF, (ii) EPA

shortened at the non-orientation (i.e., without considering orientation) of TFBSs of a TF,

and (iii) EPA without being shortened by a TFBS. EPIs predicted based on the three

types of EPAs in common were removed. EPIs predicted based on EPA (i) were

removed from EPIs predicted based on EPA (ii). EPIs predicted based on EPA (i) and

(ii) were removed from EPIs predicted based on EPA (iii). The resolution of HiChIP

chromatin interaction data was 1-5 kb; thus, EPIs were adjusted to 5 kb before

comparison. The number and ratio of EPIs overlapped with chromatin interactions were

compared two times between EPI (i) and (iii), and EPI (i) and (ii) (binomial distribution,

p-value < 0.025 for each test, two-sided, 95% confidence interval).

For comparison of EPIs with chromatin interactions (in situ Hi-C) in HMEC, the

genomic positions of chromatin interactions in the 'HMEC_4DNFI97O9SAZ.pairs.gz'

file in 4D nucleome data portal (http://data.4dnucleome.org) were converted into the

hg19   version   of   human   genome   using   the   liftOver   tool

(http://hgdownload.soe.ucsc.edu/downloads.html#liftover).   The   resolutions   of

chromatin interaction data and EPIs were adjusted to 3 kb before comparison. All

chromatin interactions were used in the comparison. For comparison of EPIs with

chromatin interactions (promoter capture Hi-C) in NPC, the resolutions of chromatin

interaction data in the 'GSE86189_npc.po.all.txt.bz2' file (GEO: GSE86189) and EPIs

were adjusted to 5 kb before comparison. All chromatin interactions were used in the

comparison.

Putative target genes for the comparison of EPIs and chromatin interactions were

selected from top 4,000 transcripts in terms of the expression level. The expression level

of putative target genes of EPIs overlapped with chromatin interactions was compared

with EPIs not overlapped with them. When the same transcriptional target gene of a TF

in an enhancer was predicted based on both an EPI overlapping a chromatin interaction

and a non-overlapping EPI, the target gene was eliminated. The distribution of

expression level of putative target genes was compared using a two-sided

Mann-Whitney test (p-value < 0.05).

Micro-C chromatin interaction data in human fibroblasts (4D nucleome data

portal ID: 4DNFICOEXGPJ) was converted into the hg19 version of the human genome

using the liftOver tool and was adjusted to 200-bp resolution. The same genomic

locations of paired chromatin interaction anchors were merged.

Correlation of expression level of gene pairs separated by TFBS

To examine the correlation of expression level of genes pairs separated by

TFBSs of a TF, gene expression data among 53 tissues were obtained from the

'GTEx_Analysis_2016-01-15_v7_RNASeQCv1.1.8_gene_median_tpm.gct.gz' file in

the GTEx database (https://gtexportal.org/home/). The correlation of log2-transformed

gene expression levels was analyzed based on RefSeq protein-coding transcripts

(RefSeq ID with 'NM_'). The same genomic locations of RefSeq transcripts were

removed, and TFBSs overlapped with RefSeq transcripts were also eliminated. Genes

closest to a TFBS and another gene were analyzed using BEDOPS closest-features

[101]. The Mann-Whitney test was performed using R.

Abbreviations

TF: Transcription factors; TFBS: Transcription factor binding sites; TSS:

Transcriptional start sites; EPA: Promoter and extended regions for enhancer-promoter

association; EPI: Enhancer-promoter interaction; FF: Forward-forward; FR:

Forward-reverse; RF: Reverse-forward; RR: Reverse-reverse; ChIP-seq:

ChIP-sequencing, chromatin immunoprecipitation followed by massively parallel DNA

sequencing; DNase-seq: DNase I hypersensitive sites sequencing; RNA-seq: RNA

sequencing; ENCODE: Encyclopedia of DNA elements;

Acknowledgements

Supercomputing resources were provided by the Human Genome Center of the Institute

of Medical Science at the University of Tokyo. Publication charges for this article were

funded by JSPS KAKENHI (Grant Number 16K00387). This work was partially

supported by JST CREST (Grant Number JPMJCR15G1), Japan.

 Figures

 Fig. 1 Chromatin interaction and enhancer-promoter association. **a** Forward–reverse

orientation of CTCF-binding sites is frequently found in chromatin interaction anchors.

CTCF can block the interaction between enhancers and promoters limiting the activity

of enhancers to certain functional domains [1-4]. **b** Computationally-defined regulatory

domains for enhancer-promoter association (EPA) [12]. The *single nearest gene*

association rule extends the regulatory domain to the midpoint between this gene's TSS

and the nearest gene's TSS both upstream and downstream. The EPA was shortened at

the forward-reverse orientation of TFBSs of a TF in this study (e.g. CTCF in the figure).

 Fig. 2 Identification of TFBSs in open chromatin regions and properties of TFBSs. **a**

 Comparison of expression level of putative transcriptional target genes. Each dot

represents the median expression level (FPKM) of target genes predicted based on

TFBSs of a TF in closed chromatin regions in promoters (x-axis) and an EPA shortened

at the FR orientation of CTCF-binding sites (y-axis) in T cells (upper left graph). The

distribution of expression level was changed according to four criteria of EPAs (four

graphs) [11]. Red dots indicate the median expression level of target genes is

significantly higher in an EPA than that in closed chromatin regions, and blue dots

indicate the median expression level of target genes is significantly lower in an EPA

than that in closed chromatin regions (Mann-Whitney test). The expression level of

target genes predicted on the basis of an EPA tend to be higher than those predicted on

the basis of closed chromatin regions, thereby implying that TFs serve as activators of

target genes. TFBSs overlapping H3K27ac ChIP-seq peaks were associated with a

significant difference of expression level of target genes, and TFBSs not overlapping

H3K27ac peaks were not associated with the difference (lower right two graphs). This

suggests that the significant difference of expression level is associated with the

enhancer (and promoter) signals of H3K27ac. **b** Distribution of TFBSs of TFs near TSS.

ZNF143 and YY1 tended to be located near TSS, and CTCF tended to be observed

more in distant regions than those around TSS in monocytes. **c** Distribution of TFBSs of

TFs in the human genome in monocytes. Most of TFBSs were found in intergenic and

intron regions. YY1 is known to be at the TSS in the edge of exons of genes and is

involved in transcriptional activation, repression, and initiation.

Fig. 3 Biased orientations of TFBSs of TFs affect the transcription of genes by the putative insulator function of the TFs. Ninety-six (sixty-four FR and fifty-two RF) biased orientations of TFBSs of TFs significantly affected the expression level of putative transcriptional target genes in monocytes, T cells, HMEC, and NPC, whereas non-orientation of TFBSs of a TF did not affect the expression level significantly.

Fig. 4 Genes separated by predicted TFBSs of biased orientations of TFs are less correlated in gene expression. The top 20 TFs are selected according to ascending order of the median correlation coefficient. The correlation coefficient between neighboring gene pairs is shown in terms of probability density **a** and cumulative distribution **b**. The brown line (All) represents the correlation between all neighboring genes; gray shading in **a** and red line (random) in **b** represents correlation between randomly chosen gene pairs; and the purple line (no_CTCF) represents the correlation between gene pairs not separated by predicted CTCF-binding sites.

Tables

Table 1 Biased orientations of DNA-binding motifs of TFs in common between monocytes, T cells, HMEC, and NPC. TF: DNA-binding motifs of transcription factors.

Forward-reverse orientation

TF	P-value	TF	P-value	TF	P-value
AHR	0	MYC	0	**ZNF143**	0
ARNT	0	MYCN	0	ZNF740	0
ARNTL	7.96E-08	MZF1	0	ZNF770	6.51E-08
ATF2	3.09E-14	NFY	1.72E-11	ZNF84	0
ATF3	2.22E-15	NFYB	0		
BCL	0	NONO	0		
BHLHE41	5.55E-14	NRF1	1.63E-11		
CREB3L1	0	OVOL2	0		
CTCF	0	PLAG1	0		
CXXC1	0	PLAGL1	0		
E2F	0	**RAD21**	0		
E2F1	0	REST	0		
EGR1	0	RFX1	0		
EGR3	0	SIN3A	0		
EGR4	0	SIX5	0		
ELF1	0	**SMC3**	0		
ELF2	5.74E-13	SP2	1.04E-11		
ELK3	3.33E-15	TFAP2A	0		
ELK4	6.19E-11	TFAP2B	2.22E-16		
ERG	0	TFAP2C	0		
ETS	0	TP73	0		
ETS1	0	USF1	0		
ETV7	2.31E-08	**YY1**	0		
HEY2	0	YY2	0		
HIC1	0	ZBTB33	0		
HINFP	6.46E-14	ZBTB7A	0		
INSM1	2.22E-16	ZFP62	0		
KLF4	0	ZFX	0		
KLF7	9.07E-09	ZIC1	0		
MAX	0				

Reverse-forward orientation

TF	P-value	TF	P-value
AP1	0	RFX2	0
ARNT	1.26E-11	RFX5	0
ATF3	0	RUNX3	0
BHLHE40	8.88E-16	SP1	0
CREB1	0	SP2	0
CTCF	0	SP4	0
E2F	2.32E-13	SPDEF	0
E2F3	0	SREBF2	0
EFNA2	6.61E-11	STAT	0
EGR1	0	TATA	0
EGR2	0	TFAP2B	2.00E-15
ETS1	0	TFAP2C	0
FLI1	1.05E-13	TFCP2L1	0
FLI1:FIGLA	0	USF1	7.55E-10
FOS	0	YBX1	0
GMEB2	0	**YY1**	0
HIF1A	0	YY2	0
KLF4	0	ZBTB33	0
MITF	0	ZBTB7B	0
MYB	0	ZNF100	6.22E-14
MYC	0	**ZNF143**	0
MYCN	0	ZNF432	0
MZF1	0		
NFKB1	0		
NFYA	2.31E-13		
NR3C1	0		
NRF1	0		
OSR2	0		
PBX3	0		
PLAG1	2.78E-14		

Table 2 Comparison of enhancer-promoter interactions (EPIs) with chromatin interactions in T cells. The upper tables show the number of biased orientations of DNA motifs, where a significantly higher ratio of EPIs that were predicted on the basis of an EPA (i) overlapped with HiChIP chromatin interactions than EPA (ii) and (iii) in T cells. The middle tables show 133 FR and 90 RF orientations of DNA motifs found in common among B2T1, B2T2, and B3T1 replications in T cells in the upper table. The lower tables show that among 64 FR and 52 RF orientations of DNA motifs found in common in monocytes, T cells, HMEC, and NPC, 44 (69%) FR and 28 (54%) RF were matched with the analysis of HiChIP chromatin interactions according to three types of EPAs. TF: DNA-binding motifs of a transcription factor. Score: $-\log_{10}$ (p-value). Score 1: Comparison of EPA (i) (an EPA shortened at the FR or RF orientation of TFBSs) with EPA (iii) (an EPA not shortened). Score 2: Comparison of EPA (i) with EPA (ii) (an EPA shortened at the non-orientation of TFBSs).

1. Comparison of EPI with HiChIP data among three EPA (i), (ii) and (iii)

Replication of HiChIP data	Total no. (FR + RF) of DNA motifs	No. of FR DNA motifs	No. of RF DNA motifs
B2T1	226	153	101
B2T2	221	149	98
B3T1	237	160	109
DNA motifs found in common among replications			
B2T1 and B2T2	212	141	95
B2T1 and B3T1	212	143	93
B2T2 and B3T1	208	138	93
B2T1, B2T2 and B3T1	201	133	90

2. Comparison of EPI with HiChIP data between two EPA (i) and (iii)

Replication of HiChIP data	Total no. (FR + RF) of DNA motifs	No. of FR DNA motifs	No. of RF DNA motifs
B2T1	395	262	205
B2T2	398	264	204
B3T1	394	263	203
DNA motifs found in common among replications			
B2T1 and B2T2	394	262	202
B2T1 and B3T1	391	261	202
B2T2 and B3T1	392	262	200
B2T1, B2T2 and B3T1	390	261	199

722   FR and RF orientations of DNA motifs found to be common among three replications in

723   T cells.

724   Forward-reverse (FR)

TF	Score 1	Score 2	TF	Score 1	Score 2	TF	Score 1	Score 2
AHR	38.81	13.41	ETS2	208.78	13.99	MNT	215.63	8.00
ARNTL	155.78	1.96	ETV1	153.32	7.02	MTF1	323.31	14.21
ATF1	323.31	15.00	ETV3	140.11	14.58	MXI1	323.31	70.98
ATF3	323.31	35.22	ETV5:EVX1	145.48	17.33	MYC	84.79	24.93
BCL	153.72	3.52	ETV5:HES7	8.76	8.74	MYCN	165.16	25.87
CREB	90.91	2.32	ETV5:TCF3	203.52	5.90	MZF1	81.83	2.29
CREB3	323.31	38.74	FOXN4	73.43	5.59	NFKB1	279.24	14.27
CREB3L1	323.31	27.96	FOXO1:ELK3	323.31	43.51	NFKB1:NFKB1	323.31	13.79
CTCF	323.31	3.50	GABPA	197.57	20.39	NFY	323.31	3.42
CTCFL	174.90	2.09	GABPA:GABPB1	142.28	6.10	NONO	323.31	9.24
CXXC1	69.92	3.30	GCM2	13.68	18.80	NR2C2	323.31	3.64
E2	323.31	20.15	GDNF	286.73	3.15	NRF1	45.13	30.75
E2F	122.12	21.87	GMEB1	136.68	10.56	OVOL2	323.31	31.61
E2F1	43.22	2.85	HIC1	223.55	4.56	PATZ1	101.85	9.30
EGR1	79.09	3.79	HIF1A	241.47	1.71	PLAG1	177.90	11.02
EGR2	176.69	4.53	HOXB2:ETV1	102.22	2.42	PURA	196.04	17.98
EGR3	85.25	34.18	ID2	323.31	14.52	R	22.49	3.99
EHF	323.31	8.66	INSM1	193.85	15.87	**RAD21**	258.21	4.81
ELF1	203.38	3.86	JUN:JUNB	275.78	6.48	RBAK	267.43	11.25
ELF2	227.62	7.83	JUNB:JUN	202.73	5.07	RELA	323.31	4.68
ELF5	323.31	8.76	KLF	252.10	8.35	RFX1	159.78	17.05
ELK1:TBX21	126.46	8.41	KLF15	323.31	22.56	RFX4	323.31	35.28
ELK1:TEF	323.31	7.99	KLF17	87.51	14.77	RFX5	323.31	26.52
ELK4	117.83	4.81	KLF2	90.74	15.34	SAP1	124.23	14.63
ENO1	80.42	3.13	KLF4	209.71	9.38	**SMC3**	222.09	1.76
EPAS1	65.43	7.76	KLF7	229.95	11.96	SP1	235.48	29.09
ERF	207.67	6.25	LHX8	244.38	2.95	SP2	94.11	9.54
ERG	199.81	25.70	MAX	323.31	8.35	SP4	14.30	6.31
ETS	196.80	1.63	MAZ	109.42	3.37	SPDEF	179.39	16.81
ETS1	308.00	3.92	MBD2	8.41	1.95	SPIC	323.31	28.49

TF	Score 1	Score 2	TF	Score 1	Score 2
STAT	84.60	1.69	ZNF322	120.49	11.55
STAT1	323.31	5.38	ZNF325	69.42	2.74
TCF12	111.44	16.79	ZNF408	214.73	2.22
TCF3	323.31	14.05	ZNF467	127.21	4.64
TEAD4:ETV4	226.28	5.90	ZNF479	226.76	3.26
TFAP2A	25.86	20.26	ZNF48	106.17	9.88
TFAP2B	90.03	6.10	ZNF511	84.74	9.87
TFAP2C	169.60	8.55	ZNF553	93.45	9.17
TFAP4	93.98	2.36	ZNF555	232.83	7.63
TFE3	139.21	16.71	ZNF624	235.50	3.84
TP73	198.45	7.75	ZNF672	266.64	10.80
TR4	39.01	4.97	ZNF675	323.31	5.81
USF1	194.06	3.56	ZNF676	323.31	16.54
VEZF1	134.36	3.49			
XBP1	323.31	44.58			
XRCC5:XRCC6	132.56	3.92			
YY1	62.59	5.30			
ZBED6	183.14	20.23			
ZBTB17	48.37	7.27			
ZBTB26	151.91	16.84			
ZBTB5	120.19	9.58			
ZBTB7A	52.89	30.58			
ZFX	13.33	14.55			
ZFY	36.57	4.04			
ZIC3	266.75	10.54			
ZIC4	17.27	9.30			
ZNF140	323.31	4.99			
ZNF143	323.31	35.38			
ZNF165	323.31	7.80			
ZNF25	281.69	7.72			

728 Reverse-forward (RF)

TF	Score 1	Score 2	TF	Score 1	Score 2	TF	Score 1	Score 2
ARNT	323.31	14.12	MAX	304.34	8.64	TFE3	95.23	10.59
ARNT2	41.41	42.78	MLXIPL	64.44	10.74	THAP1	127.72	5.07
ASCL2	138.15	6.41	MUSCLE	23.05	3.47	**YY1**	323.31	9.92
ATF3	323.31	14.64	MYB	323.31	17.68	YY2	47.49	2.61
ATF4	323.31	25.14	MYBL1	316.23	23.27	ZBTB33	90.29	7.78
ATF7	79.30	4.96	MYC	323.31	12.23	ZBTB45	323.31	57.08
BCLAF1	148.24	8.52	MZF1	200.49	22.73	ZBTB7A	89.19	16.19
BHLHE40	323.31	12.42	NFY	209.97	3.22	ZBTB7B	43.90	23.59
CDCA7L	84.72	8.42	NFYA	323.31	11.75	ZFP64	147.00	13.59
CREB3	206.55	11.07	NHLH1	105.19	4.42	ZIC3	42.65	15.39
CREB3L1	323.31	7.63	NR1H3	323.31	9.38	ZNF100	27.67	4.75
CREB3L2	67.51	7.98	NR3C1	43.16	8.31	**ZNF143**	63.03	9.69
CTCF	240.61	4.56	OSR2	111.77	14.65	ZNF148	213.77	18.93
E2F	78.16	7.07	RFX1	323.31	22.99	ZNF16	323.31	47.43
E2F1	124.10	15.73	RXRA	146.30	15.11	ZNF37A	310.67	14.79
E2F3	65.67	8.85	SALL1	277.84	1.91	ZNF410	140.45	13.57
E2F6	165.75	30.69	SCRT2	301.48	9.60	ZNF419	247.63	11.25
EFNA2	255.01	10.95	SMAD6	259.24	7.43	ZNF429	87.05	32.39
ELK3	143.48	7.57	SP100	20.75	7.99	ZNF436	323.31	11.48
EP300	121.01	10.90	SP2	294.86	2.59	ZNF468	21.12	7.67
ETV5:CLOCK	42.10	7.29	SP4	110.94	19.63	ZNF557	299.44	9.58
GLIS2	37.40	17.34	SPDEF	185.85	14.35	ZNF575	39.45	14.38
GMEB2	323.31	7.81	SPZ1	323.31	9.87	ZNF586	260.73	6.18
HES1	88.18	9.46	SREBF1	67.27	15.74	ZNF611	279.80	15.65
HES7	166.52	14.60	SREBF2	297.27	1.62	ZNF616	189.53	30.98
HIVEP1	323.31	4.90	TAF1	252.27	17.47	ZNF692	202.47	2.33
HOXB2:ELK1	217.94	11.92	TATA	2.16	4.02	ZNF727	214.01	17.22
IRC900814	323.31	4.89	TFAP2A	20.17	29.15	ZNF770	300.52	4.71
KLF4	213.35	9.49	TFAP2B	29.46	12.62	ZNF773	150.45	8.84
KLF8	323.31	17.42	TFAP2C	112.10	51.21	ZSCAN21	323.31	5.55

731 FR and RF orientations of DNA motifs found to be common in the four cell types.

732 Forward-reverse (FR)   Reverse-forward (RF)

TF	Score 1	Score 2
AHR	38.81	13.41
ARNTL	155.78	1.96
ATF3	323.31	35.22
BCL	153.72	3.52
CREB3L1	323.31	27.96
CTCF	323.31	3.50
CXXC1	69.92	3.30
E2F	122.12	21.87
E2F1	43.22	2.85
EGR1	79.09	3.79
EGR3	85.25	34.18
ELF1	203.38	3.86
ELF2	227.62	7.83
ELK4	117.83	4.81
ERG	199.81	25.70
ETS	196.80	1.63
ETS1	308.00	3.92
HIC1	223.55	4.56
INSM1	193.85	15.87
KLF4	209.71	9.38
KLF7	229.95	11.96
MAX	323.31	8.35
MYC	84.79	24.93
MYCN	165.16	25.87
MZF1	81.83	2.29
NFY	323.31	3.42
NONO	323.31	9.24
NRF1	45.13	30.75
OVOL2	323.31	31.61
PLAG1	177.90	11.02

TF	Score 1	Score 2
RAD21	258.21	4.81
RFX1	159.78	17.05
SMC3	222.09	1.76
SP2	94.11	9.54
TFAP2A	25.86	20.26
TFAP2B	90.03	6.10
TFAP2C	169.60	8.55
TP73	198.45	7.75
USF1	194.06	3.56
XBP1	323.31	44.58
YY1	62.59	5.30
ZBTB7A	52.89	30.58
ZFX	13.33	14.55
ZNF143	323.31	35.38

TF	Score 1	Score 2
ARNT	323.31	14.12
ATF3	323.31	14.64
BHLHE40	323.31	12.42
CTCF	240.61	4.56
E2F	78.16	7.07
E2F3	65.67	8.85
EFNA2	255.01	10.95
GMEB2	323.31	7.81
KLF4	213.35	9.49
MYB	323.31	17.68
MYC	323.31	12.23
MZF1	200.49	22.73
NFYA	323.31	11.75
NR3C1	43.16	8.31
OSR2	111.77	14.65
SP2	294.86	2.59
SP4	110.94	19.63
SPDEF	185.85	14.35
SREBF2	297.27	1.62
TATA	2.16	4.02
TFAP2B	29.46	12.62
TFAP2C	112.10	51.21
YY1	323.31	9.92
YY2	47.49	2.61
ZBTB33	90.29	7.78
ZBTB7B	43.90	23.59
ZNF100	27.67	4.75
ZNF143	63.03	9.69

Table 3 Correlation of the expression level of gene pairs separated by TFBSs of a TF.

Pairs of genes separated by 64 FR and 52 RF orientations of TFBSs of TFs found to be

common between monocytes, T cells, HMEC, and NPC were analyzed to find

significantly lower correlation of their expression level compared to the correlation of

all pairs of nearby genes among 53 tissues. Score: $-\log_{10}$ (p-value) of the

Mann-Whitney test. Median correlation: median correlation coefficient of expression

level of nearby gene pairs separated by TFBSs. The median correlation coefficient of

expression level of all pairs of nearby genes was 0.17. # TFBS: number of TFBSs of a

TF separating gene pairs with expression data. All TFBSs of a TF were used, except for

TFBSs separating pairs of ubiquitously expressed genes with coefficient of variance <

90.

Forward-reverse (FR)   **Reverse-forward (RF)**

TF	Score	Median correlation	# TFBS	TF	Score	Median correlation	# TFBS
AHR	1.80	0.12	520	AP1	3.87	0.11	744
ARNT	1.35	0.09	141	ARNT	2.95	0.10	453
ATF2	1.40	0.11	166	ATF3	4.34	0.10	665
ATF3	3.61	0.11	795	BHLHE40	2.49	0.07	155
BHLHE41	2.02	0.07	118	CREB1	1.79	0.11	303
CREB3L1	2.88	0.05	127	**CTCF**	5.96	0.09	721
CTCF	6.02	0.09	837	E2F	3.95	0.10	663
E2F	3.73	0.10	444	E2F3	4.05	0.09	616
E2F1	3.79	0.11	868	EFNA2	3.51	0.10	543
EGR1	2.97	0.10	391	EGR1	2.58	0.10	428
ELF1	4.86	0.10	714	EGR2	1.39	0.13	747
ELF2	3.88	0.11	781	ETS1	3.75	0.11	667
ELK3	1.86	0.12	547	FLI1:FIGLA	5.41	0.08	428
ELK4	2.44	0.11	663	FOS	3.51	0.11	777
ETS	3.34	0.11	523	GMEB2	1.36	0.09	310
ETS1	4.27	0.11	687	HIF1A	2.34	0.11	448
ETV7	3.68	0.11	662	KLF4	2.21	0.12	744
HIC1	1.65	0.11	380	MITF	3.14	0.10	419
HINFP	1.33	0.09	258	MYC	4.85	0.10	704
KLF4	2.00	0.13	718	MYCN	2.79	0.10	507
MYC	3.86	0.11	762	NFKB1	1.59	0.11	342
MYCN	2.36	0.11	583	NFYA	2.15	0.10	229
MZF1	1.74	0.10	261	NRF1	2.01	0.12	615
NFYB	2.97	0.10	264	OSR2	1.99	0.12	301
NRF1	2.21	0.11	640	PBX3	2.20	0.11	253
PLAG1	2.93	0.10	503	RFX2	2.12	0.09	168
PLAGL1	2.55	0.10	414	RFX5	3.03	0.05	75
RAD21	6.08	0.09	852	RUNX3	2.01	0.11	241
REST	4.35	0.11	920	SP1	2.73	0.11	725
RFX1	2.38	0.08	81	SP2	2.15	0.12	1090
SIN3A	3.31	0.11	813	SP4	2.33	0.11	479
SMC3	6.18	0.09	625	SPDEF	5.29	0.07	449
SP2	2.30	0.12	738	TFAP2B	3.42	0.11	855
TFAP2A	4.17	0.11	767	TFAP2C	2.05	0.11	576
TFAP2C	2.13	0.11	470	TFCP2L1	1.45	0.13	362
USF1	4.24	0.10	628	YBX1	1.34	0.13	160
XBP1	6.65	0.07	379	**YY1**	2.23	0.01	50
ZBTB33	4.90	0.10	701	ZBTB33	3.44	0.11	886
ZBTB7A	2.82	0.10	485	ZBTB7B	2.27	0.13	892
ZFP62	2.95	0.11	565	**ZNF143**	2.26	0.07	81
ZFX	2.69	0.11	581				
ZIC1	4.47	0.11	829				
ZNF770	3.38	0.08	271				

Table 4 Top 30 of the co-localized and overlapping biased orientations of DNA-binding

motifs of TFs in monocytes. Co-localized and overlapping DNA motifs of CTCF with

another biased orientation of DNA motifs are shown in separate tables. Motif 1 and

Motif 2: DNA-binding motifs of TFs. No.: the number of genomic locations with both

Motif 1 and Motif 2 upstream and downstream of genes in an EPA (< 1 Mb distance

from TSS of genes). The total number of genes was 30,523.

Forward-reverse (FR)

Motif 1	Motif 2	No.	Mean of distances bet. DNA sites (nt)	Median of distances bet. DNA sites (nt)	Mean of distances from TSS (nt)	Median of distances from TSS (nt)	Interquartile range of distance from TSS (nt)
FOS	JUN	3332	0.00	0	43202.95	18085	5524 - 48583
CTCF	**SMC3**	2876	0.01	0	34606.92	13716	3073 - 39471
FOSL2	JUN	2807	0.00	0	40389.84	17365	5424 - 45251
FOSB	FOSL2	2644	0.00	0	44302.93	17867	5633 - 47209
EGR1	EGR2	2397	0.00	0	14578.21	1251	247 - 11473
CTCF	**RAD21**	2352	0.00	0	33492.68	12913	2824 - 37672
RAD21	**SMC3**	2343	0.02	0	33572.80	14302	3856 - 37159
ATF3	FOSL2	2322	0.00	0	42784.04	17841	5366 - 48236
FOSL2	JUNB	2305	0.02	0	42884.36	17867	5706 - 46515
ETS2	ETSLIKE	2205	0.00	0	37660.66	15765	4146 - 41462
FOS	JUNB	2151	0.00	0	38403.20	16367	4849 - 44736
IRF	IRF4	1962	0.00	0	40794.41	15872	3969 - 45946
FOSL2	SMARCC1	1945	0.00	0	42429.69	17369	5094 - 46586
EGR1	WT1	1941	0.16	0	12485.23	427	93 - 5394
FOSB	JUN	1808	0.02	0	36648.74	16705	5156 - 40309
ATF3	JUN	1713	0.02	0	38101.24	17119	5294 - 43559
JUN	TFAP2A	1628	0.00	0	42262.97	18017	5089 - 47094
FOSL2	TFAP2A	1628	0.00	0	36606.38	16201	4949 - 41948
BACH2	SMARCC1	1462	0.16	0	38673.02	14778	4060 - 41262
ELF1	ETS1	1459	0.00	0	37056.72	14346	3370 - 40339
CHD2	E2F	1429	0.00	0	19220.38	1579	329 - 16257
JUN	NR3C1	1262	0.00	0	35052.17	14418	3844 - 38534
USF1	USF2	1260	0.00	0	25188.91	4726	413 - 24699
BACH2	JUNB	1238	0.15	0	36252.37	15302	4462 - 39393
BACH2	FOSB	1184	0.22	0	37190.80	14778	4038 - 41262
ETSLIKE	GABPA	1164	0.00	0	32352.24	11683	2356 - 36919
CTCF	CTCFL	1151	0.00	0	30252.04	11606	2229 - 34539
NR5A1	NR5A2	1133	0.00	0	41171.86	12243	2488 - 43136
ATF3	BACH2	1094	0.18	0	35659.16	13605	3821 - 39756
ETS2	GABPA	1092	0.06	0	27250.24	12184	2858 - 32196

760 Reverse-forward (RF)

Motif 1	Motif 2	No.	Mean of distances bet. DNA sites (nt)	Median of distances bet. DNA sites (nt)	Mean of distances from TSS (nt)	Median of distances from TSS (nt)	Interquartile range of distance from TSS (nt)
EGR1	EGR3	3012	0.00	0	17005.00	709	149 - 10393
NR1I2	NR1I3	2932	0.00	0	49622.84	20548	5924 - 58436
ZIC1	ZIC3	2871	0.07	0	21511.64	3092	434 - 19128
EGR	WT1	2612	0.04	0	9949.38	396	102 - 4266
FOS	FOS:JUN	2512	0.01	0	37200.80	15466	4438 - 42373
EGR	EGR3	2293	0.17	0	11634.92	532	135 - 6044
MAFG	NFE2L1:MAFG	2272	0.03	0	34641.56	16917	5284 - 42097
KLF10	SP1	2233	0.00	0	9778.94	297	73 - 3077
EGR	EGR2	2160	0.16	0	10502.76	490	129 - 5227
KLF3	SP3	2036	0.00	0	6974.78	246	63 - 1447
KLF1	SP3	1771	0.00	0	7748.42	264	66 - 1829
KLF10	SP3	1734	0.00	0	7275.48	222	58 - 1320
KLF7	SP4	1639	0.00	0	12521.62	309	58 - 4894
KLF1	SP1	1574	0.05	0	6780.46	245	61 - 1438
AP1	JUNB	1572	0.05	0	34158.07	15451	4746 - 40315
EGR1	WT1	1543	0.31	0	12742.28	444	111 - 5553
EGR1	SP1:SP3	1393	0.08	0	7789.54	307	91 - 1886
FOS:JUN	JUNB	1386	0.14	0	33012.80	14245	3900 - 37988
SP1	SP3	1372	0.00	0	8845.31	293	77 - 1816
EHF	ELF1	1366	0.02	0	41217.39	13357	2842 - 43178
CTCF	RXRA	1362	0.00	0	29981.33	10937	2100 - 31407
AP1	FOS	1327	0.03	0	33225.52	13455	3503 - 37399
ATF1	ATF3	1215	0.00	0	23125.06	3798	342 - 21091
ELK1	ZNF200	1156	0.05	0	22200.49	2883	181 - 17610
SP1	SP4	1137	0.00	0	7371.85	215	63 - 1232
EHF	ELF3	1134	0.00	0	36296.46	10912	596 - 37318
MEIS2	TGIF1	1018	0.00	0	47677.15	17654	3273 - 49233
KLF3	SP1	1017	0.06	0	6514.33	226	60 - 994
SPI1	ZNF200	913	0.06	0	10962.81	425	85 - 6046
AP1	FOS:JUN	860	0.09	0	50412.92	24217	7302 - 60018

Co-localized and overlapping DNA motifs of CTCF with other DNA motifs.

FR orientation of CTCF

Motif 1	Motif 2	No.	Mean of distances bet. DNA sites (nt)	Median of distances bet. DNA sites (nt)	Mean of distances from TSS (nt)	Median of distances from TSS (nt)	Interquartile range of distance from TSS (nt)
CTCF	SMC3	2876	0.01	0	34606.92	13716	3073 - 39471
CTCF	RAD21	2352	0.00	0	33492.68	12913	2824 - 37672
CTCF	CTCFL	1151	0.00	0	30252.04	11606	2229 - 34539
CTCF	ZBTB7A	470	0.27	0	23227.96	7065	745 - 23496
CTCF	E2F	123	0.81	0	7913.36	222	61 - 1997
CTCF	WT1	120	0.00	0	3543.72	114	44 - 375
CTCF	MYC	110	2.30	0	31289.45	11948	2887 - 33972

RF orientation of CTCF

Motif 1	Motif 2	No.	Mean of distances bet. DNA sites (nt)	Median of distances bet. DNA sites (nt)	Mean of distances from TSS (nt)	Median of distances from TSS (nt)	Interquartile range of distance from TSS (nt)
CTCF	RXRA	1362	0.00	0	29981.33	10937	2100 - 31407
CTCF	CTCFL	429	0.00	0	25171.22	9834	1859 - 24741
CTCF	RAD21	337	0.00	0	38829.91	17973	6117 - 45749
CTCF	ZIC1	126	1.40	0	35480.65	10286	2316 - 29380
CTCF	ZIC3	108	0.00	0	30219.32	10904	4463 - 31037

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