1   CTCF mediates the Activity-by-contact derived cis-regulatory hubs
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ABSTRACT

The 3D chromatin architecture establishes a complex network of genes and regulatory

elements necessary for transcriptomic regulation in development and disease. This network

can be modeled by cis-regulatory hubs (CRH) which underscore the local functional

interactions between enhancers and promoter regions and differ from other higher order

chromatin structures such as topological associated domains (TAD). The Activity-by-contact

(ABC) model of enhancer-promoter regulation has been recently used in the identification of

these CRHs, but little is known about the role of CTCF on the ABC scores and the consequent

impact on CRHs. Here we show that the loss of CTCF leads to a reorganization of the enhancer-

promoter interactions resulting in a re-distribution of ABC scores of the putative enhancers

in C57/bl6 mouse hearts. The loss of CTCF also leads to a global reduction of the total number

of CRHs, and an increase in the size of the CRHs due to increase in the number of elements

within each hub. In addition, CTCF loss led to more CRHs that cross TAD boundaries. These

results provide another layer of evidence to support the importance of CTCF in the formation

of regulatory networks necessary for gene regulation.

INTRODUCTION

The regulation of gene expression by the non-coding genome is an area of rapidly evolving

research globally as it is implicated in different biological phenomena including organ

development, cell-fate determination, tissue response to external stimuli and overall cellular

homeostasis [1]. Furthermore, given the localization of the majority of disease-relevant

genomic variations in non-coding regulatory regions [1], their involvement in disease

pathogenesis is well documented. Accordingly, tremendous effort is deployed to annotate

these regulatory elements in different species, tissues and cells [2-4]. To this end, different

experimental protocols and computational algorithms have been applied to identify genome-

wide regulatory elements responsible for controlling gene expression [5]. These include

techniques such as ChIP-seq, ATAC-seq, DNase-seq, Hi-C, and functional genomics using

CRISPR activation or inhibition studies etc., each with their strengths and limitations[5,6]. In

addition, computational methods such the Activity by Contact (ABC) algorithm[3], TargetFinder,

ELMER and others have been developed to integrate these datasets and identify gene-

enhancer pairs. Similarly, large consortiums have been formed to annotate these regulatory

elements in different tissues [7,8]. These studies provide a readily accessible resource and

reference catalogue of the non-coding regulatory elements to the scientific community and

help scientists unravel the role of regulatory regions in disease development and evaluate

potential therapeutic strategies [9-11].

Besides the direct identification of enhancer-gene (E-G) pairs, other studies have

demonstrated that enhancers and promoters often interact dynamically within a network

often called cis-regulatory hubs (CRH), allowing for multiple regulatory connections [12,13].

These CRHs can be defined as 3D regulatory networks that constitute a complex organization

of multi-loci connections of enhancers and promoters, connected in 3D space within the

nucleus. They highlight direct and indirect contact between genes and distal regulatory

elements and can help predict target genes for disease-relevant non-coding SNPs[13]. They are

often bound by a cluster of transcription factors and have been implicated in gene co-

regulation, lineage specification and disease development [12-14]. These hubs differ from other

known 3D structures such as genomic compartments which constitute euchromatin or

heterochromatin regions. They also differ from TADs defined as units within which most

enhancer-promoter interactions occur and loops which are pair-wise connections [15].

Identification of these CRHs is often done through detailed analysis of 3D contacts based on

Hi-C data or using chromatin accessibility (ATAC or DNAse) to study co-accessible regions and

thus construct these hubs [12,16-19]. A recent study proposed building these CRHs as a bipartite

network using the Activity-by contact (ABC) algorithm [13]. The ABC algorithm combines ATAC-

seq, ChiP-seq and Hi-C to rank the gene enhancers based on their regulatory impact and was

shown to outperform other models such as HiChIP or TargetFinder in identifying regulatory

enhancers [3,20]. Thus in addition to identifying and ranking enhancers, the ABC algorithm is

used to generate a more predictive and physiologically relevant CRH[13], and these hubs were

strongly enriched for disease-relevant genes and also helped explain disease heritability [13,21].

Given that the 3D genome architecture is a key component of the ABC algorithm, which is

used to construct the CRHs, [13], it is of huge importance to understand the role of architectural

proteins like CTCF. The CTCF protein often called the "genome weaver" plays a crucial

function in the "folding" of the genome and bringing of enhancers into proximity to their

distant target genes [15,22]. However, their role in regulating CRHs have not been explored. In

this study, using mouse cardiomyocytes as a model of terminally differentiated non-dividing

cells, we employ the ABC model to identify top cardiac gene-enhancers and then to study

CRHs. We analyse the characteristics of CRHs containing tissue-specific genes, showing that

these genes are enriched in multi-enhancer CRHs. We also show that in disease, there is a

positive regulation of genes and enhancers within a CRH. Finally we show that loss of CTCF

leads to a merging of CRHs and formation of new CRHs in the *Ctcf*-KO cells with resultant

drastic change in the number and distribution of enhancers that cross the ABC threshold.

RESULTS

Cardiac Enhancer Landscape Identified Through Activity-By-Contact Algorithm

We performed ATAC-seq, H3K27ac ChiP-seq and HiChIP , to generate the ABC scores for the

enhancer-gene pairs in the control adult mouse cardiomyocytes. Using an ABC cut-off of 0.01

[13], we identified 7000 out of 156,988 putative distal regulatory regions marked by H3K27ac

peaks and/or ATAC peak in the adult mouse CM that crossed the threshold. This represents about 5% of all putative regulatory regions predicted to have regulatory effect on their target genes and is similar to what was observed in another study [3]. With regards to the enhancer-gene (E-G) pairs, globally there were 34,496 E-G interactions with an average of 4.19 connections per gene, and 5.29 connections per enhancer in these control adult mouse CMs. Figure 1A shows examples of top-linked enhancers and their target genes for 2 cardiac disease-relevant genes *Mybpc3* and *Myh6*. The comprehensive list of ABC linked enhancers and target genes can be found in Supplementary Table 1. A literature search confirmed that at least 3 of the identified enhancers in our study have been validated. These include the *Nppa/Nppb* super enhancer located upstream of both genes and which has been shown to play a critical role in stress gene response of *Nppa/Nppb* during pressure-overload induced cardiac stress [23-25]. A separate enhancer for *Myh7* identified from the ABC scoring has also been validated previously in the mouse heart [26]. Deletion of this enhancer region led to downregulation of *Myh7* in mouse hearts[26]. Taken together, these previous studies provide support to the validity of our prediction model.

To provide further evidence that these ABC-linked enhancers regulate their putative target genes, we studied correlation between changes in gene expression versus changes in peak height of the ABC-linked enhancers during pressure overload-induced cardiac hypertrophy. Using RNA-seq and H3K27ac ChIP-seq datasets from mouse hearts subjected to transverse aortic constriction [23], we found that changes in gene expression correlates modestly with changes in H3K27ac peak signals (Pearson correlation= 0.34, *P* < 2.2e-16) (Fig 1B). The correlation was stronger when focussing only on genes and enhancers with greater or less than $\log_2$ 0.5 fold change after Transverse aortic constriction (Suppl Fig 1A) (Pearson correlation = 0.51, *P* < 2.2e-16). This result shows that globally, upon external stimuli, gene expression is regulated in the same direction as their ABC linked enhancers.

Loss Of CTCF Leads To Changes In The Enhancer Interactome And Changes In The ABC Scores Of Putative Enhancers

Next, to elucidate the role of CTCF in ABC scores and regulation of the CRHs, we also applied the ABC algorithm to the *Ctcf*-Knock out (*Ctcf*-KO) mouse cardiomyocytes. Given that previous publications have shown that the overall number of regulatory regions identified through

ChIP-seq or ATAC-seq after CTCF deletion remain mostly unchanged [27,28], with changes only

observed in the peak heights, we focussed our attention on the HiChIP differences. We have

previously published a genome-wide Hi-C in mouse cardiomyocytes [23]. A deeper analysis of

the H3K27ac interactions through the first Micro-C/MNase mediated H3K27ac HiChIP in

mouse CMs unravelled the enhancer interactome in the mouse heart. HiChIP presents

advantages over Hi-C by providing better resolution with equal or lower sequencing depth [29]

while the use of MNase in Micro-C provides better resolution for enhancer-promoter

interactions compared to restriction-enzyme digest methods [30]. The H3K27ac HiChIP

experiment was performed on two independent biological replicates each, and the data were

merged to give a total of 160 million unique pairs for the *Ctcf*-KO and 126 million unique pairs

for the control samples sufficient for HiChIP analysis[29]. For comparative analysis, the *Ctcf*-KO

data was down sampled to 126M to match the control data. Using a window size of 10kb and

an FDR of <0.1, we performed FitHiChIP [31] to identify the significant H3K27ac-anchored loops

in our dataset. The results revealed 12,843 H3K27ac loops unique to control cells, 7,070

H3K27ac loops unique to *Ctcf*-KO cells and 483 shared H3K27ac loops between both

conditions (Figure 2A, Supplementary Table 2). The KO-unique H3K27ac loops spanned longer

distances than control-unique and shared H3K27ac loops (Figure 2B), consistent with previous

studies using H3K4me3 HiChIP in zebrafish [27]. We further analysed these H3K27ac interaction

anchors for the presence of CTCF using the Control CTCF ChIP-seq as reference. We

discovered that 14% of these H3K27ac loop anchors occurred in regions with CTCF binding at

both anchors in the control-unique group, while only 4.2% of the H3K27ac loops in the *Ctcf*-

KO occurred in anchors with native CTCF binding. Concordantly, 56% of the H3K27ac loops

had no native CTCF binding at the anchors in control, with the number increasing to 65% in

the KO samples. The percentage of H3K27ac loops with native CTCF binding at one anchor

was consistent at about 30% for both control and *Ctcf*-KO (Fig 2C). This shows that the loss of

CTCF led to a 70% decrease in the H3K27ac loops that co-localize with CTCF at both anchors

in the *Ctc*f-KO cells, with an increase in ectopic interactions between H3K27ac anchors

without CTCF. Next we performed the aggregate peak analysis (APA) for the control and *Ctcf*-

KO samples, the APA reveals enrichment for loop domains often bound by CTCF and Cohesin

at the anchors [15]. Consistent with our previously published Hi-C data, and other studies that

have performed aggregate analysis in different *Ctcf*-KO models [32], there was a marked

decrease in the APA score from 3.33 in the control to 0.86 in the *Ctcf*-KO for all H3K27ac loops,

suggesting indeed the loss or weakening of loop domains in the *Ctcf*-KO samples [23].The APA

score was also higher for the control unique and shared loops in the control samples, while

the APA score for *Ctcf*-KO unique group was higher in the *Ctcf*-KO samples as anticipated,

possible representing some re-arrangement of Cohesin in the *Ctcf*-KO group (Figure 2D)

We applied the ABC algorithm on the *Ctcf*-KO sample and observed that there were 36,226

regulatory elements that crossed the 0.01 threshold, resulting in 71,080 E-G interactions with

similar number of expressed genes (~8,000) (Supplementary Table 1) in control and *Ctcf*-KO

samples. This led to an increase in the average number of connections per gene to 8.14 in

Ctcf-KO samples. Of the 71,080 enhancer-gene pairs only 20% (14,200 E-G pairs) were shared

between control and Ctcf-KO. This increase in ABC scores could be due to increased

interaction frequency and/or increased H3K27ac ChIP-seq upon CTCF loss. Interestingly we

observed that the highest ranked ABC enhancer in the control had a score of 0.99, while the

highest ranked enhancer in the *Ctcf*-KO sample had a score of 0.78 (scale of 0 – 1). Indeed the

average score in the control was 0.054 while the average score in the *Ctcf*-KO was 0.028 (TTest

P-value < 2.2e-16) . This suggests that the loss of CTCF led to gain of ectopic enhancers in the

Ctcf-KO with a redistribution of the ABC scores of the enhancers.

Next we integrated the ABC results to generate the CRHs focussing first on the control

samples. We identified 1,522 hubs in these control cardiomyocytes with about 70% of the

hubs containing fewer than 5 elements (Fig 3A). Figure 3B shows 2 examples of such hubs

containing cardiac genes: *Myom1* a gene involved in formation of myofibrils [33], and found in

a simple hub while *Tnni3* a cardiac-specific sarcomeric gene is found in a more complex hub.

Consistent with the correlation between gene expression and ABC-linked enhancers in heart

disease, we also verified that genes and enhancers within the same hubs are positively

correlated in same direction of change during cardiac stress, showing Pearson correlation of

0.32 (P <2.2e-16) (Figure 3C). This suggests a co-regulation of genes and enhancers found

within the same hub upon external stimuli. We then ranked cardiac genes by FPKM and

selected the top 50 genes, mostly cardiac-specific genes, to glean biological insights about the

characteristics of CRHs that harbour these highly expressed genes. Our data revealed that

highly expressed cardiac genes were more likely to be in hubs with more regulatory elements

than non-highly expressed genes (Figure 3D). These cardiac-specific genes were also more

likely to be found in hubs with higher number of promoters. Such multi-enhancer hubs has

been proposed as a mechanism to buffer against environmental stresses and genetic

perturbations and thus provide phenotypic robustness to key genes [21,34]. On the other hand,

hubs with high number of promoters are characteristic of lineage specific genes [13,21].

Loss Of CTCF Alters The Cis-Regulatory Hubs

Finally, we analysed the CRHs in the *Ctcf*-KO CMs and found a marked reduction in total

number of CRHs from 1522 in control to 660 CRHs in the KO, accompanied with an increase

in the number of elements per hub. Fig 4A shows example of the same hubs containing 2

cardiac genes shown in Fig 3A. Describe observation., Figure 4B shows the distribution of

elements per hub, confirming a striking increase in the percentage of hubs with 10 or more

elements. Associated with this increase in percentage of complex hubs, there was an increase

in the average number of connections per gene in the *Ctcf*-KO cells, and a decrease in number

of connections per enhancer (Fig 4C, Supplementary Fig 1B). We asked if this reduction in

total number of CRHs was merely due to merging of CRHs or formation of new CRHs, and

observed both cases. The merging of CRHs was observed primarily in the simple hubs (2-3

elements per hub) located within close proximity as 403 (82%) of 492 such simple hubs

merged with 1 or more other hubs to form larger hubs. The proportion of the number of CRHs

merging to form a new one varied, ranging from 2 merged CRHs to 15 merged CRHs

Supplementary Fig 1C. For the larger hubs, there was mostly formation of new hubs, as genes

gained new interaction and lost other interactions. To further confirm the importance of CTCF

in the organization of these hubs, we compared CTCF binding sites in the control hubs vs *Ctcf*-

KO hubs based on control CTCF ChIP-seq, and observed an enrichment of CTCF in the control

hubs (fisher exact test odds ratio of 1.69, two sided $P < 2e\text{-}16$). Next, we analyzed the

relationship between CRHs and TADs since TAD boundaries are enriched for CTCF [23], and

earlier studies have shown that CRHs are generally constrained within TAD boundaries [13].

Indeed, using the TADs in the control as reference, our data showed that about 80% of CRHs

in the control were contained within the same TAD while only 20% spanned more than 1 TAD.

In contrast, about 45% of CRHs in KO were contained within a TAD while 50% of CRHs in the

KO spanned more than 1 TAD (Fig 4D) while 5% were not within TAD boundaries, showing

that the loss of CTCF leads to a reorganization of the CRHs with the formation of cross-TAD

enhancer-promoter interactions.

DISCUSSION

Non-coding regulatory elements play crucial roles in development and disease and hold
promise for next-generation therapeutic targets. While different studies by us and others
have annotated these elements and CRHs in different cellular models, the role of CTCF in the
ABC derived CRHs was yet to be studied. First, our study applied the ABC algorithm to the
mouse heart, generating to the best of our knowledge, the first such ranked enhancer scores
in mice hearts. Our findings confirm the positive correlation between genes and their ABC-
linked enhancers in cardiac disease, and affirms that genes and enhancers within the same
hub are co-regulated during pressure overload-induced cardiac hypertrophy. Integrating ABC
to identify CRHs may thus represents another method to analyze for effect of regulatory SNPs
on target genes, as the SNP may have effect on a target gene when they are in the same hub,
even when there is no direct link through pair-wise enhancer-promoter analysis [13]. Analysis
of the CRHs also suggests that tissue-specific genes are more likely to be contained in hubs
with high connectivity and rich in distal elements. This increased number of E-P networks
provides redundancy and phenotypic robustness while guaranteeing increased
transcriptional activity thanks to the multi-enhancer networks.

Loss of CTCF affects the enhancer interactome, particularly impacting the H3K27ac
interactions co-bound by CTCF. This leads to E-Ps that span larger distances and gain of new
interactions. The loss of CTCF also leads to a re-organization of ABC-selected enhancers with
a direct impact on the CRHs, resulting in fewer hubs but with more elements per hub. This
implies a crucial role for CTCF in not only determining which enhancers cross the ABC
threshold but also in the organization of the ABC-derived CRH. While this study focussed on
the effect of loss of CTCF on overall CRH structure and its relationship with TAD boundaries,
future studies will examine individual CRHs and their relationship to CTCF binding to ascertain
how loss of CTCF can lead to local dysregulation of CRHs. Indeed, this finding has implications
for mutations that affect local CTCF binding at regions that are not TAD boundaries, as it may
lead to merging of 2 or more CRHs and thus the establishment of indirect connection between
genes and enhancers within the newly formed hubs.

Our findings therefore demonstrate that CTCF plays a critical role in determining the
membership of the CRHs as well as their organization, and thus plays a role in disease and cell
type-specific gene expression through organization of these cis-regulatory hubs.
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ABBREVIATIONS

300 ChIP-seq – Chromatin immunoprecipitation with sequencing

301 ATAC-seq – Assay for Transposase Accessible Chromatin with sequencing

302 Hi-C – High throughput chromatin conformation capture with sequencing

303 ABC – Activity by contact algorithm

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DECLARATIONS

ETHICS APPROVAL: Animal protocol approved by the Institutional Animal Care and Use
Committee (National University of Singapore)

CONSENT FOR PUBLICATION – The authors consent to publication of this article

AVAILABILITY OF DATA AND MATERIALS

The raw data supporting the conclusion of this article will be made available by the authors,
without undue reservation.

COMPETING INTEREST

Cory C. Padilla works for Cantata Bio, developers of the HiChIP kit used in the study. The other
authors declare that the research was conducted in the absence of any commercial or
financial relationships that could be construed as a potential conflict of interest

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AUTHOR CONTRIBUTION

M.L, D.P.L, W.Z, L.G, C.C.K, Y.L, R.W, C.G.A.N performed the experiments, M.L, L.M, C.P, W.T,
S.B, A.B, C.G.A.N provided data analysis. S.B, A.B, R.F, C.G.A.N Provided supervision for
experiments and data analysis. All authors read and approved the manuscript

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METHODS

Animal Experiments

Animal experiments were performed under a license approved by the Institutional Animal

Care and Use Committee (National University of Singapore). *Ctcf*[flox/flox] mice harboring LoxP

sites flanking exons 3 and 12 of the *Ctcf* gene against a C57/bl6 strain background were used

as previously published [23]. Details of the source of the original strain has been documented

in our earlier publication [23]. AAV9-cTnt-eGFP and AAV9-cTnt-Cre-tdTomato vectors encoding

codon improved Cre recombinase or eGFP under the control of cardiac troponin T promoter

were used for control and *Ctcf* knock-out respectively [23]. Experiments were performed on

adult 6- to 8-week-old mice. 3 mice were used for each group, housed in 2 different cages.

The numbers were based on our previously published study [23]. Cardiomyocytes were

isolated from mouse hearts after 2 weeks of AAV9 injections for control and *Ctcf*-KO mice.

CTCF knock-out was assessed by western blot, and upon confirmation, cells were used for

subsequent experiments. Adeno-associated virus 9 (AAV9) virus for the delivery of Cre

recombinase was generated as previously described [23]. For cardiomyocyte isolation,

isoflurane anaesthesia is used before performing a thoracotomy to harvest the hearts as

previously described [23].

H3k27ac HiChip Library Prep And Data Analysis.

Mouse adult cardiomyocytes were isolated following previously published protocol [23].

MNase-based HiChIP assay was performed on 2×10^6 isolated CMs. Frozen cells were

resuspended in 1× PBS and crosslinked with 3 mM DSG and 1% formaldehyde. Washed cells

were digested with 0.75 µl MNase in 100 µl of nuclease digest buffer with $MgCl_2$. Cells were

lysed with 1× RIPA, and clarified lysate was used for ChIP. H3K27ac antibody (Abcam 04729)

was used to perform chromatin immunoprecipitation. The Protein A/G bead pulldown,

proximity ligation, and libraries were prepared as described in the Dovetail protocol (Dovetail

HiChIP MNase Kit). Libraries were sequenced on an Illumina Novaseq.

HiChIP paired-end reads were aligned with BWA MEM (version 0.7.17r1198-dirty) with the -

5SP flag enabled with an index containing only the canonical chromosomes of the mm10

genome (available from the UCSC genome). The resulting alignments were then parsed with

pairtools (versions 0.3.0) with the following options --min-mapq 40 –walks-policy 5unque –max-inter-align gap 30 and the thre –chroms-path file corresponding to the size of the chromosomes used for the alignment index. Parse pairs were deduplicated, sorted with pairtools. Valid pairs were identified and through pairtools select (pair type = 'UU', 'UR', 'RU', 'uu', 'uU' and 'Uu') and downsampled to the lowest number of valid pairs in a sample (126 million in LACZ) with pairtools sample. Contact matrices in .hic format were generated with juicetools pre function (version 1.22.01).

FitHiChIP (version 9.1) was used to identify significant interactions ("loops") from the valid subsampled pairs at 10kb resolution with the following settings, loop type = all-to-all, coverage = bias correction, merge redundant loops = Yes, background model = FithHiChIP(L), FDR < 0.1, minimum interaction size =20kb, maximum interaction size = 2 mb. Conditionally unique (meaning only occurring in LACZ or CRE) and shared loops were identified with bedtools pairToPair (version 2.28.0) requiring both loop anchors to overlap the same coordinates be flagged as shared. Loop size was assessed by performing a two-sided Wilcoxson's rank-sum test on the distribution of loop ranges (distance between the two anchors). Loop anchors were then intersected the LACZ CTCF binding sites to determine the proportion of loops resulting from CTCF presence. P-values for CTCF mediated loops were obtained through a two-sided Fischer's Exact test. Aggregate Peak Analysis (APA) was computed with juicetools APA function. Loop anchors were used to as the sites to aggregate over at 10kb resolution. APA enrichment scores, loop center from the lower-left (LL) corner is shown as both an APA score (ratio of the mean of center to mean of LL) and a Z-score. APA normed output was used to plot the APA matrices.

H3K27ac HiChIP data were summarized and visualized with R, the package eulurr was used for the shared and unique loop Venn diagrams and ggplot2 for loop size and proportion of loops with CTCF binding site at the anchor. Additionally, ggplot2 was used to plot APA matrices with the geom_raster function, with the color scale set to the same min – max limits across all APA plots. At sites of interest, HiChIP loops files (in longrange format) were used to visualize loops along with ChIP-seq coverage and RNA-seq activity in the WashU Epigenome Browser (https://epigenomegateway.wustl.edu/).

Chip-Seq Experiment And Data Analysis

ChIP experiments on isolated CM were performed as described previously [23]. Briefly, CMs

were cross-linked with 1% formaldehyde for 10 minutes at room temperature and quenched

with glycine (0.125 mol/L final concentration). Cells were then washed in ice-cold PBS and

pelleted and lysed in FA lysis buffer (10 mmol/L Tris-HCl, pH 8.0, 0.25% Triton X-100, 10

mmol/L EDTA, 0.1 mol/L NaCl). To facilitate cell lysis, the cell pellet was passed through a 27.5-

gauge needle gently 20 times. Nuclei were pelleted by centrifugation resuspended in

sonication buffer, and chromatin was fragmented via sonication to an average size of 200 to

300 bp (Bioruptor Plus, Diagenode). Chromatin was immunoprecipitated against H3K27ac

(Abcam ab4729) or CTCF (EMD Millipore, catalog No. 07–729) overnight. Antibody-chromatin

complexes were pulled down with Protein G Dynabeads (Invitrogen, catalog No. 10003D),

washed with 0.1% SDS lysis buffer, and eluted with elution buffer (1% SDS, 10 mmol/L EDTA,

50 mmol/L Tris-HCl, pH 8). After cross-link reversal (4 hours of incubation at 65°C) and

proteinase K treatment, immunoprecipitated DNA was extracted. ChIP DNA was quantified

by fluorometric quantification (Thermo Fisher Scientific, Qubit dsDNA HS assay kit, catalog

No. 32851). Library preparation was performed with the New England Biolabs Ultra II Kit

according to the manufacturer's specifications and sequenced on the Illumina NextSeq High

platform.

Paired-end reads were aligned to the mm10 genome using Bowtie2. The uniquely aligned

reads were PCR-duplicate removed. Peak-calling was performed using Dfilter, and

parameters were optimized for different antibodies. For H3K27ac, the parameters are: ks=60,

lpval=6. For CTCF, ks=10, lpval=6, -nonzero, -refine. Reads were counted in 100-bp windows

for each library and scaled to normalize for sequencing depth. Read counts in each window

were then adjusted by normalizing their GCcontent against the average GC-content of all

libraries. Peak height was obtained by summing the bin-wise normalized counts. Peak height

was quantile-normalized to reduce technical variation. Supplementary Table 3 lists H3K27ac

peaks in Sham and TAC treated mouse cardiomyocytes.

ATAC-Seq Experiment And Analysis

The ATAC-seq was performed according to previously published Omni-ATAC protocol [35]

Briefly, 50,000 adult cardiomyocytes were pelleted at 500 g for 5 minutes. The cells were

resuspended in 100 µl ATAC-resuspension buffer containing 0.5% NP40, 0.5 % Tween-20 and

0.01% Digitonin. The cells were left on ice for 5 minutes after which 1 ml of cold ATAC-RSB

containing 0.1% Tween 20 was added to wash out the lysis buffer. The nuclei were pelleted

at 500 g for 5 minutes, the supernatant was carefully removed and the nuclei were

resuspended in 50 µl transposition mixture containing 25 µl of 2x TD buffer, 16.5 µl of PBS, 5

µl nuclease free water, 0.5 µl of 1% digitonin, 0.5 µl of 10% Tween-20 and 2.5 µl of transposase

(Illumina Tagment DNA enzyme 1, Catalogue number 20034198). The reaction was incubated

at 37°C for 30 minutes in a thermoshaker with 1000 RPM. After the reaction, DNA was

extracted using the NEB Monarch® PCR & DNA cleanup kit (Catalogue number T1030L), PCR

was performed using the Illumina/Nextera primers after which Ampure XP beads were used

for library clean up. The resulting library was sequenced on Nextseq. Note that while ChIP-

seq and HiChIP were performed on control and CTCF-KO cells, the ATAC-seq experiments

were performed on a wild type C57/bl6 strain mice. ATACseq peaks were called on each

sample using MACS2 (callpeak function with these parameters: -nomodel and -B). BedGraph

files generated by MACS2 where normalized to read counts per million reads sequenced

ABC-Score And CRH Analysis

The ABC model defines active enhancers based on a quantitative score of DNAse, H3K27ac,

and normalized Hi-C contact number[3]. This score is computed relative to a background activity

over a 5-Mb window around a candidate element. Then, we set the threshold to 0.01; beyond

which a candidate element is considered as a distal element. As an extension of the ABC-

Score, CRHs were defined as bipartite networks between promoters and distal elements

(igraph R package) [36] TADs were called using the directionality index as previously described.

Pearson correlation analysis performed using R package

Figure 1:

A) USCS screenshot showing 2 examples of cardiac genes and their top ABC-linked

enhancers

B) Pearson analysis showing correlation between differential expressed genes and
differential enhancer peaks for ABC-linked enhancers during cardiac disease
455

Figure 2: H3K27ac HiChIP in Control and CTCF KO
A) Venn diagram showing number of FitHiChIP interactions in Control-unique and *Ctcf*-
KO unique and shared loops
B) Box plot showing the loop sizes in the various groups. The loops in CTCF-unique
interactions span longer distance than the interactions in Control unique CM
C) Bar chart showing distribution of CTCF at the FitHiChIP anchors. 14% of the loops have
CTCF at both anchors in the control while only 4.2% of loops in the CTCF KO have CTCF
on both anchors using the Control CTCF biding as reference
D) Aggregate Peak Analysis for the loops in Control and CTCF KO cells, showing a marked
reduction of the APA and Z-score in the CTCF KO samples for all loops
466

Figure 3: CRH in control Cardiomyocytes
A) Bar chart showing distribution of number of elements per CRH
B) Examples of 2 CRHs containing cardiac genes *Myom1* and *Tnni3* in the control
cardiomyocytes
C) Pearson correlation between differential expression of genes and enhancer peaks
contained within the same CRHs during cardiac stress
D) Box plot showing characteristics of CRHs containing cardiac specific genes. These CRHs
tend to have more enhancers and promoters
475

Figure 4: CRHs in the KO cells
A) Examples of 2 CRHs containing cardiac genes *Myom1* and *Tnni3* in the *Ctcf*-KO
cardiomyocytes. Compared to the same CRHs in control, there is an increase in the
number of elements in each hub
B) Bar chart showing distribution of elements per CRH in Control and *Ctcf*-KO. There is
an increase in the percentage of CRHs with 10 or more elements from 15% in the
Control to 60% in the CTCF KO cardiomyocytes

C) Bar chart showing number of connections per gene promoters and per enhancers.
484   There is an increase in the average number of connections per gene in the *Ctcf*-KO
485   CMs
486   D) Bar chart showing overlap of CRHs with TADs, 80% of TADs are contained in one TAD
487   in the control while only 40% of CRHs are contained within one TAD in the CTCF KO.
488   In contrast 20% of CRHs span 2 or more TADs in the control, while 50% of CRHs span
489   2 or more TADs in CTCF KO.
490

491 Supplementary figure 1
492   A) Pearson analysis showing correlation between differential expressed genes and
493   differential enhancer peaks for ABC-linked enhancers during cardiac disease when
494   taking into account only genes and enhancers with a fold change of log2 0.5
495   B) Bar chart showing number of connections per gene promoters and per enhancers in
496   control and *Ctcf*-KO cells
497   C) Bar chart showing how many CRHs were merged to form a bigger CRH in *Ctcf*-KO cells,
498   and the percentage of each category
499

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600

A

Top
Enhancer

Mybpc3
gene

Myh6
gene

Top
Enhancer

B

$$y = -0.067 + 0.13\,x$$
$$R = 0.34,\ p < 2.2e\text{-}16$$

Log$_2$FC_FPKM

Log$_2$FC_H3K27ac Peaks

A Loop Summary

Control Unique 12,843 — Shared 483 — CTCF KO Unique 7,070

All Loops 20,396

B Loop Size

P < 0.00001
P < 0.00001   P < 0.00001

Loop Size (kbp): 0, 100, 200, 300, 400, 500

Control Unique   Shared   CTCF KO Unique

C Loop Anchors Occuring Over CTCF

% of Loops: 0, 20, 40, 60, 80, 100

Anchros with no CTCF
One Anchor with CTCF
Both Anchors with CTCF

P < 0.00001
P = 0.268   P < 0.00001

Control Unique   Shared   CTCF KO Unique

D APAs

All Loops | Control Unique | Shared | CTCF KO Unique

Control

APA LL = 3.33, Z-score LL = 82.04
APA LL = 3.33, Z-score LL = 82.05
APA LL = 2.75, Z-score LL = 11.35
APA LL = 0.88, Z-score LL = 2.23

CTCF KO

APA LL = 0.86, Z-score LL = -3.88
APA LL = 0.87, Z-score LL = -3.89
APA LL = 2.00, Z-score LL = 7.54
APA LL = 2.23, Z-score LL = 29.16

-100kb Loop +100kb
-100kb Loop +100kb
-100kb Loop +100kb
-100kb Loop +100kb

A

B

C

Supplementary Figure