Data in Brief

Genome-wide profiling of YY1 binding sites during skeletal myogenesis

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Abstract

Skeletal muscle differentiation is regulated by a network of transcription factors, epigenetic regulators and non-coding RNAs. We have recently performed ChIP-seq experiments to explore the genome-wide binding of transcription factor YY1 in skeletal muscle cells. Our results identified thousands of YY1 binding peaks, underscoring its multifaceted functions in muscle cells. In particular, we identified a very high proportion of YY1 binding peaks residing in the intergenic regions, which led to the discovery of some novel lincRNAs under YY1 regulation. Here we describe the details of the ChIP-seq experiments and data analysis procedures associated with the study published by Lu et al. in the EMBO Journal in 2013 [1].

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Direct link to deposited data

http://www.ncbi.nlm.nih.gov/geo/query/acc.cgi?acc=GSE45875

Experimental Design, Materials and Methods

Cell culture

Mouse C2C12 myoblast cell line was purchased from American Type Culture Collection (ATCC). The myoblasts were maintained in a growth medium (DMEM, 10% FBS and 1% penicillin/streptomycin), and induced to myotubes by culturing in a differentiation medium (DMEM, 2% horse serum and 1% penicillin/streptomycin).

ChIP assays and sequencing experiments

ChIP assays were performed as previously described [2,3]. About $2 \times 10^7$ C2C12 cells and 5 μg of antibodies were used in one immunoprecipitation. The antibodies include YY1 #1 (Santa Cruz Biotechnology, Cat# SC-1703, rabbit polyclonal), YY1 #2 (Abcam, Cat# AB58066, mouse monoclonal), Ezh2 (Cell Signaling, MA, USA, Cat# AC22), trimethyl-histone H3-K27 (Millipore, Cat# 07-449), trimethyl-histone H3-K4 (Millipore, Cat# 07-473), or normal mouse IgG (Santa Cruz Biotechnology, Cat# SC-2025) as a negative control.

For library construction, we used a protocol as described before [4]. Briefly, the immunoprecipitated DNA (~10 ng) were end-repaired, and A-nucleotide overhangs were then added, followed by adapter ligation, PCR enrichment, size selection and purification. The purified DNA library products were evaluated using Bioanalyzer (Agilent) and SYBR qPCR and diluted to 10 nM for sequencing on Illumina Hi-seq 2000 sequencer (YY1) (pair-end with 50 bp) or Illumina Genome Analyzer II sequencer (Ezh2, H3K27me3 and H3K4me3) (pair-end with 36 bp). Technical replicates were prepared by sequencing the same library twice. A data analysis pipeline CASAVA 1.8 (Illumina) was employed to perform the initial bioinformatic analysis (base calling). Table 1 lists all the experiments that we had performed. For MB YY1, we performed two biological replicates with the antibody SC-1703 and a third biological replicate with a second antibody AB58066. We also performed two technical replicates for each antibody (run 1 and run 2).
The sequenced reads were mapped to the mouse reference genome (UCSC mm9, non-repeat-masked) using SOAP2 [5] (version 2.2.0, with the following parameters: "-v -r -0 -m 0 -p 20") allowing a maximum of two mismatches and only the uniquely aligned reads were kept. The protein–DNA binding peaks were identified using Model-based Analysis for ChiP-seq (MACS [6], version 2.0.9; for YY1 ChiP-seq (MB rep1); the parameters are "-g mm -m 8.30 -p 0.001" and then the peaks were filtered by q-values; for others, the parameters were "-g mm -m 8.30 -q 0.01") with the IgG control sample as background. During the peak calling, a q-value (adjusted P-value calculated using the Benjamini–Hochberg procedure) was set under $10^{-4}$ for YY1; it corresponds to an empirical FDR (False Discovery Rate) of $3.4\%$. The peaks were filtered by FDR (False Discovery Rate) of $3.4\%$. For other experiments, two peaks were considered as "overlapped peaks" if the distance between them was less than 1 kb. In order to search for highly occurring motifs in the DNA sequences underlying the putative binding peaks, Discriminative Regular Expression Motif Elicitation (DREME [7], version 4.8.0) was applied on the 100 bp ($\pm$ 50 bp) sequences flanking the peak summit. The analysis was run on both strands to search for motifs that are no more than 8 bp in length with E-values $<0.01$.

Quality control

In peak defining, we used the IgG as a negative control and also carefully selected the q-values for a reasonable FDR. According to the ENCODE ChiP-seq guidelines [8], we calculated the Fragments in Peaks (FRIP) value using in-house programs (See Supplementary Material). Moreover, for the YY1 biological replicates, we performed Irreproducible Discover Rate (IDR) analysis using the package developed by Li et al. [9].

### Functional annotation

To identify putative YY1 target gene, each identified peak was associated with the closest RefSeq gene when it falls into the 4 kb ($\pm 2$ kb) flanking region of the gene's TSS, and these genes were considered as potentially regulated by YY1. For analysis of differentially expressed genes, we used Cufflinks [10] (version 1.3) to evaluate the expression profile (using Fragments Per Kilobase of exon model per Million mapped reads, FPKM) of all the RefSeq transcripts using the publically available RNA-Seq data obtained from $-24$ h (myoblasts, MBs) and 60 h (myotubes, MTs) C2C12 [11]. Differentially expressed genes were defined as those up- or down-regulated in MTs as compared to MB. If a gene is differentially expressed and bound by YY1, we reason that it could be potentially regulated by YY1 since the YY1 level decreases during C2C12 differentiation. Up-regulated YY1 bound genes were defined if their expression in MTs is $>1.2$ fold higher compared with MBs and these genes are possibly repressed by YY1 in MBs. Down-regulated YY1 bound genes were defined if their expression in MTs is less than 0.8 fold compared with MBs and they are likely activated by YY1 in MBs. Then Gene Ontology (GO) analysis was performed on both up- and down-regulated genes using Database for Annotation, Visualization and Integrated Discovery (DAVID, http://david.abcc.ncifcrf.gov/) [12,13] for functional annotations.

### Identification of YY1 bound novel lincRNAs

Since we observed that more than 1/4 of the YY1 peaks were in the intergenic regions, we suspected that YY1 may regulate unannotated lincRNAs. To validate this hypothesis, we used the list of novel lincRNA identified by Guttman et al. from four mouse cell types [14]. YY1-binding sites were searched in the flanking regions ($\pm 100$ kb on both sides) of these lincRNAs. The resultant list of lincRNAs was considered as YY1-associated muscle lincRNAs or Yams.

### Conflict of interest statement

The authors declare no conflict of interest.

### Appendix A. Supplementary data

Supplementary data to this article can be found online at http://dx.doi.org/10.1016/j.gdata.2014.05.008.

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