SEA: a super-enhancer archive

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ABSTRACT

Super-enhancers are large clusters of transcriptional enhancers regarded as having essential roles in driving the expression of genes that control cell identity during development and tumorigenesis. The construction of a genome-wide super-enhancer database is urgently needed to better understand super-enhancer-directed gene expression regulation for a given biology process. Here, we present a specifically designed web-accessible database, Super-Enhancer Archive (SEA, http://sea.edbc.org). SEA focuses on integrating super-enhancers in multiple species and annotating their potential roles in the regulation of cell identity gene expression. The current release of SEA incorporates 83,996 super-enhancers computationally or experimentally identified in 134 cell types/tissues/diseases, including human (75,439, three of which were experimentally identified), mouse (58,795, five of which were experimentally identified), Drosophila melanogaster (1774) and Caenorhabditis elegans (904). To facilitate data extraction, SEA supports multiple search options, including species, genome location, gene name, cell type/tissue and super-enhancer name. The response provides detailed (epi)genetic information, incorporating cell type specificity, nearby genes, transcriptional factor binding sites, CRISPR/Cas9 target sites, evolutionary conservation, SNPs, H3K27ac, DNA methylation, gene expression and TF ChIP-seq data. Moreover, analytical tools and a genome browser were developed for users to explore super-enhancers and their roles in defining cell identity and disease processes in depth.

INTRODUCTION

Super-enhancers are genome regions that are large clusters of transcriptional enhancers (1). The term ‘super-enhancer’ was used for the first time by Chen et al. in 2004 (2). Super-enhancers differ from typical enhancers in size, transcription factor density and content, and the ability to activate transcription (3). As key gene regulators, super-enhancers are assembled in a cell type-specific manner, which is closely related to the binding of numerous transcription factors and the enrichment of H3K27ac (4,5).

Super-enhancers are regarded as the regulators of cell identity, stem cell pluripotency and even tumor pathogenesis (6–8). They have been reported to drive the cell type-specific expression of genes that define cell identity during development and disease (9–10), as shown by the super-enhancers of hair follicle stem cells that drove reporter gene expression in vivo (11). In another study, the CRISPR/Cas9 genome editing strategy revealed that a super-enhancer was responsible for over 90% of Sox2 expression in mouse embryonic stem cells (12). Super-enhancers play important roles in normal development, although their aberrant assembly can enhance the abnormal expression of key genes and contribute to multiple diseases including Alzheimer’s disease, type 1 diabetes mellitus and cancers (13). For example, the quintessential oncogene MYC is regulated by super-enhancers in most cancers (1,14–16), while the super-enhancers associated with genes encoding the transcription factors RUNX1, FOSL2 and BHLHE40 are critical for the mesenchymal transformation of brain tumors (17).

Several studies have identified novel cancer therapeutics directed at components of super-enhancers in diverse tumor types (18). For example, tumor oncogenes including MYC can be selectively inhibited by disrupting super-enhancers in cancers (18,19), while CDK7 inhibition suppresses super-enhancer-linked oncogenic transcription (20–22). The deletion of super-enhancer constituents using a CRISPR/Cas9-based approach in colorectal cancer cells previously reduced the expression of the associated gene, suggesting that...
cancer therapeutics targeting super-enhancer components may be particularly effective in tumor cells (9).

At present, there are several methods for discovering super-enhancers, including computational method based on factors such as Med1 (19), BRD4 (10) and H3K27ac (1), and experimental technology such as ChIA-PET (8), CRISPR/Cas9 (9), RT-qPCR (23) and so on. Because of the larger size of super-enhancers, it was difficult and time consuming to experimentally identify novel super-enhancers from a range of cell types/tissues/diseases (12). It is known that H3K27ac can identify a large fraction of super-enhancers formed by the master transcription factors and it has been profiled in a broad range of samples from multiple species (4). Thus, H3K27ac can be used as the surrogate mark of super-enhancers and imported into ROSE software developed by Loven et al. (19) to identify the super-enhancers in a broad range of samples from multiple species. Briefly, it stitches constituent enhancers together if they are within a certain distance and ranks the enhancers by their input-subtracted signal of H3K27ac. Then super-enhancers are separated from typical enhancers by identifying a threshold point of H3K27ac signal versus enhancer rank (4). For example, H3K27ac was used to identify the super-enhancers in five human DL-BCL lines and a normal tonsil sample (3). Zhou et al. also used H3K27ac signals to characteristic Epstein–Barr virus super-enhancers (5). These results indicate that H3K27ac is a robust marker of super-enhancers in various cell types, normal tissues and diseases. Recently, the development of several epigenome projects, such as ENCODE (24), mod-ENCODE (25) and Human Epigenome Roadmap (26), have provided a genome-wide landscape of H3K27ac in many cell types/tissues/diseases. This raised the possibility of integrating these data sets to systematically identify a comprehensive map of super-enhancers in multiple species.

Therefore, the comprehensive collection and more and more prediction of candidate super-enhancers would be useful for the systemic exploration of their potential roles in human disease or biology processes (14,27). To this end, we developed a specifically designed database, Super-Enhancer Archive (SEA, http://sea.edbc.org), which focuses on integrating experimentally and computationally identified super-enhancers and annotating their potential roles in the regulation of cell identity gene expression in a cell type-specific manner. Recently, another database for super-enhancers dbSUPER at http://bioinfo.au.tsinghua.edu.cn/dbsuper/ was proposed by Khan et al. (28). In contrast to dbSUPER, SEA (i) currently stores integrated super-enhancers of four species, including humans, mice, Drosophila melanogaster and Caenorhabditis elegans; (ii) displays the cell type/disease specificity of super-enhancers by comparative analysis, which is useful for the analysis of the super-enhancer regulatory function; (iii) identifies related genes of super-enhancers based on the genome location shown in flexible online networks; (iv) identifies CRISPR/Cas9 target sites of super-enhancers, which is useful for further functional analysis of super-enhancers for that the Cas9 nucleases can be directed by short RNAs to induce precise cleavage at endogenous genomic loci in organism cells; (v) supplements the classification of publications related to super-enhancers according to research content; (vi) offers an online analytical tool for genomic region enrichment analysis and another for H3K27ac cell type specificity analysis; and (vii) develops a genome browser and supports customized and user-friendly views on a genomic scale. In brief, as a comprehensive super-enhancer resource, SEA provides a convenient platform to store, annotate and readily analyse super-enhancers.

OVERVIEW AND DATA PROCESSING

In this study, we took advantage of the available public H3K27ac data sets and the published super-enhancer lists to construct a comprehensive list of super-enhancers in multiple cell types/tissues/diseases of four species (Figure 1). First, we examined the super-enhancers identified by Hnisz et al. in 86 human cell types/tissues using the ROSE program and based on the H3K27ac ChIP-Seq data set (1). Second, for other cell types/tissues in humans and other species, we obtained the H3K27ac raw data set profiled by ChIP-Seq from the SRA database (29). Bowtie2 (version 2.2.5) (30) was used to map ChIP-Seq reads to reference genomes downloaded from UCSC Genome Bioinformatics (31), including those of human (hg19), mouse (mm9), D. melanogaster (dm6) and C. elegans (ce10). For each sample, ROSE was used to obtain potential super-enhancers as ‘python ROSE.main.py -g hg19 -i /data/Human.gff -r /data/SRR*******_sort.bam -o /data/SRR******* -s 12500’. And it was run with a stitching distance of 12 500 bp, i.e. enhancers were allowed to be stitched together within 12 500 bp. Besides, we manually collected eight super-enhancers that had been identified and confirmed by experimental methods in published papers.

In this way, we incorporated 83 996 super-enhancers that had been experimentally discovered or computationally identified in 134 cell types, tissues, or diseases from humans, mice, D. melanogaster and C. elegans (Table 1). In the case of humans, there were 75 439 super-enhancers from 99 cell types/tissues including 30 disease cell lines/tissues. Among these super-enhancers, 58 283 were predicted by Hnisz et al. in 86 human cell types/tissues based on the H3K27ac ChIP-Seq data set (1), 1109 were predicted by Loven et al. in three human cell types/tissues based on the Med1 ChIP-Seq data set (19), 629 were predicted by Kwiatkowski et al. in the human Jurkat cell based on the H3K27ac ChIP-Seq data set (21), 15 415 were predicted by SEA in 15 human cell types/tissues based on the H3K27ac ChIP-Seq data set and 3 had been confirmed experimentally (3,14). In the case of mice, there were 5879 super-enhancers from 21 cell types/tissues, including 5 that had been experimentally confirmed (8,9,12). For D. melanogaster, there were 1774 computationally identified super-enhancers from 11 tissues, and 904 computationally identified super-enhancers from three tissues of C. elegans. The detailed number of samples for each cell type/tissue/disease per species that used to identify super-enhancers by different researchers were listed in the ‘Supplement Table’.

On the convenience of using of super-enhancers, its’ IDs/names in SEA are coded using a specific convention: <reference genome>_cell-
Figure 1. Database content and construction. SEA took advantage of the available public H3K27ac data sets to identify super-enhancers in different cell types/tissues/diseases of four species. And it also stores genetic and epigenetic information related to super-enhancers. Users can input multiple options to the query engine to acquire the super-enhancers. It enable users to do genomic region enrichment analysis and cell type specificity analysis. All search and analysis results can be downloaded as flat format for further analysis.

Table 1. SEA data content and statistics as of September 24, 2015

| Type            | Total num. | Data source       | Num. | Types num. | Human | Mouse | D. melanogaster | C. elegans |
|-----------------|------------|-------------------|------|------------|-------|-------|-----------------|------------|
| Super-Enhancer  | 83,996     | Hnisz et al. (1)  | 58,283| 86         | 58,283| 0     | 0               | 0          |
|                 |            | Loven et al. (19) | 1,109| 3          | 1,109 | 0     | 0               | 0          |
|                 |            | Kwiatkowski et al. (21) | 629 | 1          | 629   | 0     | 0               | 0          |
|                 |            | SEA (c)           | 23,967| 50         | 15,415| 5874 | 1774            | 904        |
|                 |            | SEA (e)           | 8     | 4          | 5     | 0     | 0               | 0          |
|                 |            | Hnisz et al. (1)  | 122  | 98         | 122   | 0     | 0               | 0          |
|                 |            | Kwiatkowski et al. (21) | 48 | 21         | 48    | 0     | 0               | 0          |
| H3K27ac         | 196        | SRA (29)          | 122  | 98         | 122   | 0     | 0               | 0          |
|                 |            | GEO (37)          | 48   | 21         | 48    | 0     | 0               | 0          |
|                 |            | modENCODE (25)    | 26   | 14         | 0     | 15    | 11              |            |
| DNA Methylation | 26         | Roadmap (26)      | 26   | 26         | 21    | 5     | 0               | 0          |
| Expression      | 35         | Roadmap (26)      | 26   | 26         | 21    | 5     | 0               | 0          |
| Reference Genome| 26         | ENCODE (38)       | 15   | 15         | 15    | 0     | 0               | 0          |
| CpG Islands     | 4          | UCSC (31)         | 4    | -          | hg19  | mm9   | dm6             | cc10       |
| SNP             | 66,916     | UCSC (31)         | 66,916| -          | 28,691| 16,026| 22,199          | 0          |
|                 | 14,730,558 | GWASdb (39)       | 314,237| -          | 314,237| 0     | 0               | 0          |
|                 |            | OMIM (40)         | 148  | -          | 148   | 0     | 0               | 0          |
|                 |            | UCSC (31)         | 14,416,173| -        | 14,416,173| 0     | 0               | 0          |
| TFBS            | 1,104,318  | UCSC (31)         | 1,104,318| -        | 893,489| 267,980| 13,357          | 19,492     |
| TF ChIP-seq data| 98         | ENCODE (38)       | 98   | 11         | 98    | 0     | 0               | 0          |
| CRISPR/Cas9     | 1,211,142  | CRISPR Genome     | 1,211,142| -        | 518,266| 137,529| 482,150         | 73,197     |

(c) Computationally identified super-enhancers by ROSE.
(e) Experimentally identified super-enhancers from literature search.

In addition, SEA also stores genetic and epigenetic information related to super-enhancers, including sequence conservation, information about associated genes, CRISPR/Cas9 target sites, H3K27ac ChIP-seq data, DNA methylation, RNA-seq data, disease and common single-nucleotide polymorphisms, transcriptional factor binding sites and TF ChIP-seq data (Table 1).

The genes associated with experimentally discovered super-enhancer were defined by the context of published literature, while the genes associated with predicted super-
enhancers were assigned by its' closest gene referred to the description in Kwiatkowski et al.’s paper. In order to simplify the back-end database, the H3K27ac ChIP-seq data, DNA methylation, RNA-seq data and TF ChIP-seq data were pre-processed following these procedures: (i) extend to upstream and downstream in the center of the super-enhancer regions with 2-fold of their original length; (ii) merge the extended regions to larger regions; and (iii) calculate the methylation level or counts of ChIP-seq/RNA-seq reads falling within per 100 bp in the merged regions, which is used as the estimation of the signal of methylation/H3K27ac/TF /expression. Then, these data sets were designed as browser view tracks for visualization.

For each TF, the enrichment score were calculated following this formula below:

\[
enrichment\ score = \frac{N(TFBS \text{ of } TF \text{ in } SE) \times L(ref\ genome)}{L(SE) \times N(TFBS \text{ of } TF \text{ in } ref\ genome)}
\]

where \(N(TFBS \text{ of } TF \text{ in } SE)\) represents the number of TFBS referred to specific TF in the super-enhancer, \(L(ref\ genome)\) represents the length of the reference genome, \(L(SE)\) represents the length of super-enhancer and \(N(TFBS \text{ of } TS \text{ in } ref\ genome)\) represents the number of TFBS referred to the tested TF in the whole reference genome. Briefly, the enrichment score is used to depict the intensity of transcription factor binding sites along the super-enhancer regions. And it is suggested that the high enrichment score may be a possible active super-enhancer.

### DATABASE USE AND ACCESS

#### Using the search tool to retrieve super-enhancers

The search tool supports accurate searching (Figure 2). Users can specify detailed information about super-enhancers such as species, genome location, gene name, cell type/tissue, super-enhancer name and transcription factor. The search tool results include genomic location, super-type enhancers such as species, genome location, gene name, cell type and (iii) calculate the methylation level or counts of ChIP-seq/RNA-seq reads falling within per 100 bp in the merged regions, which is used as the estimation of the signal of methylation/H3K27ac/TF /expression. Then, these data sets were designed as browser view tracks for visualization.

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where \(N(TFBS \text{ of } TF \text{ in } SE)\) represents the number of TFBS referred to specific TF in the super-enhancer, \(L(ref\ genome)\) represents the length of the reference genome, \(L(SE)\) represents the length of super-enhancer and \(N(TFBS \text{ of } TS \text{ in } ref\ genome)\) represents the number of TFBS referred to the tested TF in the whole reference genome. Briefly, the enrichment score is used to depict the intensity of transcription factor binding sites along the super-enhancer regions. And it is suggested that the high enrichment score may be a possible active super-enhancer.

#### Using the search tool to retrieve super-enhancers

The search tool supports accurate searching (Figure 2). Users can specify detailed information about super-enhancers such as species, genome location, gene name, cell type/tissue, super-enhancer name and transcription factor. The search tool results include genomic location, super-enhancer ID coded by SEA, associated gene, cell type/tissue, super-enhancer name and transcription factor. The search tool results include genomic location, super-enhancer ID coded by SEA, associated gene, cell type specificity, sequence conservation and CRISPR/Cas9 target sites (Figure 2A). Users can obtain detailed information about a particular super-enhancer following the links appended in this table, while information about the means used to define the super-enhancer, its' reporter, nearby genes and transcription factors are given on the detailed results page (Figure 2B). Moreover, the comparison between cell types/tissues is shown in a data matrix. Links to the SEA-browser enable visualization of the information in a genomic context (Figure 2C). The search result also provides useful links to other gene annotation databases including NCBI Gene (32), GeneCards (33) and UniProt (34). In addition, the statistic page provides statistics charts (Figure 2D) and the download page (Figure 2E) provides the download links for all data sources used in SEA.

#### Online analysis tools for super-enhancers

Using the advanced search for super-enhancers or the interface to build the bed file on the ‘Data Analysis’ page, users can obtain a ‘.bed’ file (Figure 2F), which can be downloaded for future study or directly sent to the Genomic Regions Enrichment of Annotations Tool (GREAT) (35), for the functional significance analysis of these super-enhancers. Users can also submit genomic region(s) to SEA through the second tab control on the ‘Data Analysis’ page, which may be stored in a ‘.bed’ file. Then, the submitted genomic regions were converted to super-enhancers’ regions through back-end overlapping analysis. Next, the average histone modification level of H3K27ac in the overlapped super-enhancer regions were calculated for each selected cell type/tissue, with which the heat map (Figure 2G) was generated by the JavaScript tool ‘FusionCharts’. The row and the column in the heat map represent a particular super-enhancer and a special cell type/tissue, respectively. In order to demonstrate the difference between super enhancer and other genome regions such as typical enhancer, the enrichment score calculating tool for TFs was embedded into the third tab control on the Data Analysis page. The background of the enrichment analysis is the whole genome or the genome regions given in a custom designed BED file by user. Then, SEA will calculate the enrichment score for each TF falling into the region of the specified super-enhancers.

#### SEA-browser

The SEA-browser was developed to visualize super-enhancer information in a (epi)genomic context (Figure 2C). Users can observe the proximity of super-enhancers to nearby genes, transcription factor binding sites with their ChIP-seq peaks, and cell type-specific super-enhancers, CRISPR/Cas9 target sites, H3K27ac modification pattern, DNA methylation level, expression of RNA-seq data, as well as disease single-nucleotide polymorphisms. Users can also add or reduce data tracks by clicking the selector at the bottom of the page. In addition, a ‘Custom Track’ tool has been developed to allow users to upload their own sequencing data as well as visualization.

#### A case application of SEA in the systematic analysis of super-enhancers close to human gene SOX2

On the Search page, we selected ‘Human’ as the species and input ‘SOX2’ as the gene name (Figure 3A). After clicking the ‘Search’ button at the bottom, the search result page listed eight super-enhancers related to human SOX2 in different cell types/tissues (Figure 3B), which enriched in nine molecular functions analyzed by GREAT (Figure 3C). Because SOX2 is a key gene in stem cells (36), we selected the super-enhancer (SEID: 52416) in human stem cell H1. Clicking the blue-colored SEID (52416) revealed details of this super-enhancer on an information page (Figure 3D). This showed the location of two genes (SOX2 and SOX2-OT) near the super-enhancer. The network and tables suggested the presence of five transcription factors with eight binding sites (including the well-known super-enhancer markers CEBP and MYOD) within this super-enhancer, of which some were essential for the transcription regulation of SOX2.

To examine the relationship between the super-enhancer and SOX2, we navigated to the SEA-Browser via the links provided on the search result page or the detailed informa-
Figure 2. A comprehensive view of SEA query results. (A) The search tool results table, including SEID, super-enhancer’s name, genomic location, associated gene, region length, cell types/tissues and conservative score. (B) Details of a selected super-enhancer from the results table showing the means used to define the super-enhancer, its reporter, the nearby genes and transcription factors, together with the comparison among cell/tissue types. (C) Visualization of genetic and epigenetic information in the SEA-browser, such as H3K27ac modification pattern, DNA methylation level, expression of RNA-seq and so on. (D) Quantitative statistics of super-enhancers and the sources in SEA. (E) Data sources listed in the download table. (F) Genomic region enrichment analysis tool for super-enhancers. (G) Heat map result for the cell type specificity analysis of H3K27ac modification pattern. Each of the row and the column in the heat map represented a particular super-enhancer and a special cell type/tissue, respectively.

SYSTEM DESIGN AND IMPLEMENTATION

SEA was constructed based on three major software components: an Apache Tomcat web server, a MySQL relational database and Java-based computational services. The backstage processing programs were written in Java, which are available on request. The web services were developed using Apache Struts2, a Java web application framework, and iBATIS, a persistence framework that automates the mapping between MySQL databases and objects in Java, both of which help guarantee the high performance and stability of the web services. Browser-based interfaces were built in HTML/JSP and AJAX. The Apache Batik SVG toolkit was used to render, generate and manipulate the SVG dynamically. IE 9.0+ and Firefox with 1440*900 (or higher) are preferred to achieve the best display effect. SEA is available at http://sea.edbc.org.

FUTURE DEVELOPMENT

To build a super-enhancer database focusing on multi-species, continued efforts will be made to update the SEA data and improve the SEA-browser and database func-
Figure 3. The usage of SEA by the example gene SOX2. (A) Input ‘Human’ as the species and ‘SOX2’ as the gene name in the search form. (B) The search results related to human SOX2 in different cell types/tissues. (C) Enrichment results of the eight super-enhancers from the search results analyzed by GREAT. (D) Details of the super-enhancer (SEID: 52416) in human stem cell H1 on an information page. (E) Visualization of human gene SOX2 and super-enhancers nearby.
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tionality. In next years, genome-wide super-enhancer information will be becoming accumulating faster for many cell types/tissues/diseases in more species, we will continuously collect the latest data sets to keep SEA up-to-date. Currently, nearly all super-enhancers in SEA were predicted by computational method based on H3K27ac. The super-enhancers identified or confirmed by experiment would be recorded into SEA. And more genetic and epigenetic information related to super-enhancer would be added into this database for further study. We expect that the continuous efforts to develop and improve SEA will contribute to our understanding of super-enhancers and their roles in defining cell identity and disease.

SUPPLEMENTARY DATA

Supplementary Data are available at NAR Online.

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