

# CexoR : An R package to uncover high-resolution protein-DNA interactions in ChIP-exo replicates

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## 1 Citation

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Madrigal P (2015) CexoR: an R/Bioconductor package to uncover high-resolution protein-DNA interactions in ChIP-exo replicates. **EMBnet.journal** 21, e837. <http://dx.doi.org/10.14806/ej.21.0.837>.

## 2 Introduction

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For its unprecedented level of resolution, chromatin immunoprecipitation combined with lambda exonuclease digestion followed by sequencing (ChIP-exo) is a potential candidate to replace ChIP-seq as the standard approach for high-confidence mapping of protein-DNA interactions. Numerous algorithms have been developed for peak calling in ChIP-seq data. However, adjusting the statistical models to ChIP-exo making use of its strand-specificity can improve the identification of protein-DNA binding sites. The midpoint between the strand-specific paired peaks formed at its forward and reverse strands is delimited by the exonuclease stop sites, within the protein binding event is located (Rhee and Pugh, 2011).

### 3 Methodology

Lambda exonuclease stop site (5' end of the reads) counts are calculated separately for both DNA strands from the alignment files in BAM format using the Bioconductor [Rsamtools](#). Counts are then normalized using linear scaling to the same sample depth of the smaller dataset. Using the Skellam distribution (Skellam, 1946), **CexoR** models at each nucleotide position the discrete signed difference of two Poisson counts at forward and reverse strands, respectively. Then, detecting nearby located significant count differences of opposed sign (peak-pairs) at both strands allows **CexoR** to delimit the flanks of the protein binding event location at base pair resolution. A one-sided  $p$ -value is obtained for each peak using the complementary cumulative Skellam distribution function, and a final  $p$ -value for the peak-pair (default cut-off  $1e - 12$ ) is reported as the sum of the two  $p$ -values. To account for the reproducibility of replicated peak-pairs, which central point must be located at a user-defined maximum distance,  $p$ -values are submitted for irreproducible discovery rate estimation (Li et al., 2011). Stouffer's and Fisher's combined  $p$ -values are given for the final peak-pair calls. Finally, BED files containing reproducible binding event locations formed within peak-pairs are reported, as well as their midpoints.

More information can be found in Madrigal (2015).

### 4 Example

We downloaded the 3 replicates of human CTCF ChIP-exo data from GEO (SRA044886) (Rhee and Pugh, 2011), and aligned the reads to the human reference genome (hg19) using Bowtie 1.0.0. Reads not mapping uniquely were discarded. We can search reproducible binding events between peak-pairs in the first million bp of Chr2 in the 3 biological replicates by:

```
R> options(width=60)
R> ## hg19. chr2:1-1,000,000
R>
R> owd <- setwd(tempdir())
R> library(CexoR)
R> rep1 <- "CTCF_rep1_chr2_1-1e6.bam"
R> rep2 <- "CTCF_rep2_chr2_1-1e6.bam"
R> rep3 <- "CTCF_rep3_chr2_1-1e6.bam"
R> r1 <- system.file("extdata", rep1, package="CexoR", mustWork = TRUE)
R> r2 <- system.file("extdata", rep2, package="CexoR", mustWork = TRUE)
R> r3 <- system.file("extdata", rep3, package="CexoR", mustWork = TRUE)
R> peak_pairs <- cexor(bam=c(r1,r2,r3), chrN="chr2", chrL=1e6, idr=0.01, N=3e4, p=1e-12)
R> peak_pairs$bindingEvents
```

GRanges object with 13 ranges and 6 metadata columns:

	seqnames	ranges	strand	IDR
	<Rle>	<IRanges>	<Rle>	<numeric>
[1]	chr2	11501-11701	*	0
[2]	chr2	18785-18886	*	0
[3]	chr2	142184-142371	*	0
[4]	chr2	172170-172354	*	0
[5]	chr2	332699-332870	*	0
...	...	...	...	...
[9]	chr2	662610-662783	*	0
[10]	chr2	667465-667634	*	0

```
[11] chr2 714362-714545 * | 0
[12] chr2 715918-716096 * | 0
[13] chr2 850211-850402 * | 0
      rep1.neg.log10pvalue rep2.neg.log10pvalue
      <numeric>          <numeric>
[1]      30.4163          23.8800
[2]      17.1723          23.6215
[3]      14.0090          14.1269
[4]      17.1729          17.3140
[5]      13.7083          20.3044
...      ...          ...
[9]      30.4174          34.1026
[10]     30.4163          44.8251
[11]     27.0319          20.6463
[12]     17.1725          30.6523
[13]     30.1158          23.9693
      rep3.neg.log10pvalue Stouffer.pvalue Fisher.pvalue
      <numeric>          <numeric>    <numeric>
[1]      23.5194           0           0
[2]      23.5195           0           0
[3]      20.0154           0           0
[4]      23.5195           0           0
[5]      13.9014           0           0
...      ...          ...      ...
[9]      26.8376           0           0
[10]     27.0159           0           0
[11]     26.6166           0           0
[12]     23.6777           0           0
[13]     13.9016           0           0
-----
seqinfo: 1 sequence from an unspecified genome
```

```
R> peak_pairs$bindingCentres
```

GRanges object with 13 ranges and 6 metadata columns:

```
      seqnames      ranges strand |      IDR
      <Rle>      <IRanges> <Rle> | <numeric>
[1] chr2 11601-11602 * | 0
[2] chr2 18836-18837 * | 0
[3] chr2 142278-142279 * | 0
[4] chr2 172262-172263 * | 0
[5] chr2 332784-332785 * | 0
...      ...      ...      ...
[9] chr2 662696-662697 * | 0
[10] chr2 667550-667551 * | 0
[11] chr2 714454-714455 * | 0
[12] chr2 716007-716008 * | 0
[13] chr2 850306-850307 * | 0
      rep1.neg.log10pvalue rep2.neg.log10pvalue
      <numeric>          <numeric>
[1]      30.4163          23.8800
[2]      17.1723          23.6215
[3]      14.0090          14.1269
[4]      17.1729          17.3140
[5]      13.7083          20.3044
...      ...          ...
```

```

[9]          30.4174          34.1026
[10]          30.4163          44.8251
[11]          27.0319          20.6463
[12]          17.1725          30.6523
[13]          30.1158          23.9693
      rep3.neg.log10pvalue Stouffer.pvalue Fisher.pvalue
      <numeric>          <numeric>          <numeric>
[1]          23.5194              0              0
[2]          23.5195              0              0
[3]          20.0154              0              0
[4]          23.5195              0              0
[5]          13.9014              0              0
...          ...              ...              ...
[9]          26.8376              0              0
[10]          27.0159              0              0
[11]          26.6166              0              0
[12]          23.6777              0              0
[13]          13.9016              0              0
-----
seqinfo: 1 sequence from an unspecified genome

R> setwd(owd)
R>

```

13 reproducible peak-pair events are reported for the established thresholds ( $p\text{-value} \leq 1e-12$ ,  $IDR \leq 0.01$ ). We can now plot the mean profile of lambda exonuclease stop sites and reads, 500 bp around the central position of reproducible peak-pair locations, by running the function "plotcexor":

```

R> options(width=60)
R> plotcexor(bam=c(r1,r2,r3), peaks=peak_pairs, EXT=500)

```

#### Important notes:

- For the correct estimation of the IDR (Li et al., 2011) peak-pair calling should be relaxed (e.g.,  $p\text{-value}=1e-3$ , or smaller depending on the sequencing depth), enabling the noise component be present in the data and therefore allowing the peak-pairs to be separated into a reproducible and an irreproducible groups. In the example shown above, as the dataset is very small and peaks are highly reproducible, IDR in the overlapped peak-pairs across the 3 replicates is zero.
- IDR calculation could produce varying results depending on the choice of initial estimates for four parameters needed by the algorithm ( $\mu$ ,  $\sigma$ ,  $\rho$ ,  $\text{prop}$ ). Li et al. (2011) recommend trying several choices, so that the parameter estimation does not get trapped in a local maximum.
- For more information about using IDR in high-throughput sequencing datasets see Land et al. (2012) and Bailey et al. (2013), or the mathematical description in Li et al. (2011).

## 5 References

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- Madrigal P (2015) CexoR: an R/Bioconductor package to uncover high-resolution protein-DNA interactions in ChIP-exo replicates. **EMBnet.journal** 21: e837.
- Bailey TL, et al. (2013). Practical Guidelines for the Comprehensive Analysis of ChIP-seq Data. **PLoS Comput Biol** 9: e1003326.
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- Skellam JG (1946) The frequency distribution of the difference between two Poisson variates belonging to different populations. **J R Stat Soc Ser A** 109: 296.
- Li Q, Brown J, Huang H, Bickel P (2011) Measuring reproducibility of high-throughput experiments. **Ann Appl Stat** 5: 1752-1779.
- Rhee HS, Pugh BF (2011) Comprehensive genome-wide protein-DNA interactions detected at single-nucleotide resolution. **Cell** 147: 1408-1419.

## 6 References

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## 7 Details

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This document was written using:

```
R> sessionInfo()

R Under development (unstable) (2025-03-01 r87860 ucrt)
Platform: x86_64-w64-mingw32/x64
Running under: Windows Server 2022 x64 (build 20348)

Matrix products: default
  LAPACK version 3.12.0

locale:
 [1] LC_COLLATE=C
 [2] LC_CTYPE=English_United States.utf8
 [3] LC_MONETARY=English_United States.utf8
 [4] LC_NUMERIC=C
 [5] LC_TIME=English_United States.utf8

time zone: America/New_York
tzcode source: internal

attached base packages:
[1] stats4      stats      graphics  grDevices  utils
[6] datasets   methods   base

other attached packages:
[1] CexoR_1.45.0      IRanges_2.41.3
```

```
[3] S4Vectors_0.45.4      BiocGenerics_0.53.6
[5] generics_0.1.3

loaded via a namespace (and not attached):
 [1] SummarizedExperiment_1.37.0 impute_1.81.0
 [3] gtable_0.3.6                rjson_0.2.23
 [5] xfun_0.51                   ggplot2_3.5.1
 [7] Biobase_2.67.0              lattice_0.22-6
 [9] seqPattern_1.39.0           tzdb_0.5.0
[11] vctrs_0.6.5                 tools_4.5.0
[13] bitops_1.0-9                curl_6.2.1
[15] parallel_4.5.0              tibble_3.2.1
[17] pkgconfig_2.0.3             KernSmooth_2.23-26
[19] Matrix_1.7-3                data.table_1.17.0
[21] BSgenome_1.75.1             RColorBrewer_1.1-3
[23] lifecycle_1.0.4             GenomeInfoDbData_1.2.14
[25] stringr_1.5.1               compiler_4.5.0
[27] idr_1.3                     Rsamtools_2.23.1
[29] Biostrings_2.75.4           munsell_0.5.1
[31] BiocStyle_2.35.0            codetools_0.2-20
[33] GenomeInfoDb_1.43.4         htmltools_0.5.8.1
[35] RCurl_1.98-1.16             yaml_2.3.10
[37] pillar_1.10.1               crayon_1.5.3
[39] BiocParallel_1.41.2         DelayedArray_0.33.6
[41] abind_1.4-8                 tidyselect_1.2.1
[43] digest_0.6.37               stringi_1.8.4
[45] reshape2_1.4.4              dplyr_1.1.4
[47] genomation_1.39.0          restfulr_0.0.15
[49] fastmap_1.2.0               grid_4.5.0
[51] colorspace_2.1-1            cli_3.6.4
[53] SparseArray_1.7.6           magrittr_2.0.3
[55] S4Arrays_1.7.3              XML_3.99-0.18
[57] readr_2.1.5                 UCSC.utils_1.3.1
[59] scales_1.3.0                plotrix_3.8-4
[61] rmarkdown_2.29              XVector_0.47.2
[63] httr_1.4.7                  matrixStats_1.5.0
[65] hms_1.1.3                   evaluate_1.0.3
[67] knitr_1.50                  GenomicRanges_1.59.1
[69] BiocIO_1.17.1               rtracklayer_1.67.1
[71] rlang_1.1.5                 Rcpp_1.0.14
[73] gridBase_0.4-7              glue_1.8.0
[75] BiocManager_1.30.25         jsonlite_1.9.1
[77] plyr_1.8.9                  R6_2.6.1
[79] MatrixGenerics_1.19.1       GenomicAlignments_1.43.0
```