

The ChIPpeakAnno user's guide

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Contents

1	Introduction	2
2	Quick start	2
3	Examples of using ChIPpeakAnno	3
3.1	Task 1: Find the nearest feature such as gene and the distance to the feature such as the transcription start site (TSS) of the nearest gene	4
3.2	Task 2: Obtain overlapping peaks for potential transcription factor complex and determine the significance of the overlapping and generate Venn Diagram	7
3.3	Task 3: Obtain sequences surrounding the peaks for PCR validation or motif discovery	12
3.4	Task 4: Obtain enriched gene ontology (GO) terms or KEGG terms near the peaks	13
3.5	Task 5: Find peaks with bi-directional promoters	15
3.6	Task 6: Output a summary of motif occurrence in the peaks.	16
3.7	Task 7: Add other IDs to annotated peaks or enrichedGO	16
3.8	Task 8: annotate ChIP results from BED or GFF files or MACS output xls file .	17
4	References	21
5	Session Info	22

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1 Introduction

Chromatin immunoprecipitation (ChIP) followed by high-throughput tag sequencing (ChIP-seq) and ChIP followed by genome tiling array analysis (ChIP-chip) become more and more prevalent high throughput technologies for identifying the binding sites of DNA-binding proteins in a genome-wide bases. A number of algorithms have been published to facilitate the identification of the binding sites of the DNA-binding proteins of interest. The identified binding sites in the list of peaks are usually converted to BED or WIG file format to be loaded to UCSC genome browser as custom tracks for investigators to view the proximity to various genomic features such as genes, exons and conserved elements. However, clicking through the genome browser could be a daunting task for the biologist if the number of peaks gets large or the peaks spread widely across the genome.

Here we have developed a Bioconductor package called `ChIPpeakAnno` to facilitate the batch annotation of the peaks identified from either ChIP-seq or ChIP-chip experiments. We have implemented functionality to find the nearest gene, exon, miRNA, gene end or custom features supplied by users such as most conserved elements and other transcription factor binding sites leveraging `IRanges`. Since the genome annotation gets updated from time to time, we have leveraged the `biomaRt` package from Bioconductor to retrieve the annotation data on the fly if the annotation of interest is available via the `biomaRt` package. The users also have the flexibility to pass their own annotation data as `GRanges` (or `RangedData`) or pass in annotation data from `GenomicFeatures`. We have also leveraged `BSgenome` and `biomaRt` package on implementing functions to retrieve the sequences around the peak identified for peak validation. To understand whether the identified peaks are enriched around genes with certain GO terms, we have implemented GO enrichment test in `ChIPpeakAnno` package leveraging the hypergeometric test phyper in `stats` package and integrated with Gene Ontology (GO) annotation from `GO.db` package and multiplicity adjustment functions from `multtest` package.

2 Quick start

```
> library(ChIPpeakAnno)
> ## import the MACS output
> macs <- system.file("extdata", "MACS_peaks.xls", package="ChIPpeakAnno")
> macsOutput <- toGRanges(macs, format="MACS")
> ## annotate the peaks with ensembl annotation
> data(TSS.human.GRCh38)
> macs.anno <- annotatePeakInBatch(macsOutput, AnnotationData=TSS.human.GRCh38,
+                               output="overlapping", maxgap=5000L)
> ## add gene symbols
> library(org.Hs.eg.db)
> macs.anno <- addGeneIDs(annotatedPeak=macs.anno,
+                        orgAnn="org.Hs.eg.db",
+                        IDs2Add="symbol")
> head(macs.anno)
```

`GRanges` object with 6 ranges and 16 metadata columns:

```

      seqnames          ranges strand | length summit tags
      <Rle>            <IRanges> <Rle> | <factor> <factor> <factor>
X01.ENSEG00000117616 chr1 [ 25323511, 25324015] * | 505 252 45
X01.ENSEG00000187010 chr1 [ 25323511, 25324015] * | 505 252 45
X02.ENSEG00000183726 chr1 [ 25362685, 25362997] * | 313 211 33
X02.ENSEG00000188672 chr1 [ 25362685, 25362997] * | 313 211 33
      X03.NA chr1 [145558152, 145558537] * | 386 59 39
      X04.NA chr10 [ 47088702, 47089329] * | 628 484 68
      qvalue fold_enrichment FDR peak feature
      <factor> <factor> <factor> <character> <character>
X01.ENSEG00000117616 59.17 17.01 5.8 X01 ENSG00000117616
X01.ENSEG00000187010 59.17 17.01 5.8 X01 ENSG00000187010
X02.ENSEG00000183726 60.63 22.41 4.2 X02 ENSG00000183726
X02.ENSEG00000188672 60.63 22.41 4.2 X02 ENSG00000188672
      X03.NA 53.10 20.68 2.3 X03 <NA>
      X04.NA 56.09 16.37 0.75 X04 <NA>
      start_position end_position feature_strand insideFeature
      <integer> <integer> <character> <factor>
X01.ENSEG00000117616 25242237 25338213 - inside
X01.ENSEG00000187010 25272393 25330445 + inside
X02.ENSEG00000183726 25337917 25362361 + downstream
X02.ENSEG00000188672 25362249 25430192 - inside
      X03.NA <NA> <NA> <NA> <NA>
      X04.NA <NA> <NA> <NA> <NA>
      distancetoFeature shortestDistance fromOverlappingOrNearest
      <numeric> <integer> <character>
X01.ENSEG00000117616 14702 14198 Overlapping
X01.ENSEG00000187010 51118 6430 Overlapping
X02.ENSEG00000183726 24768 324 Overlapping
X02.ENSEG00000188672 67507 436 Overlapping
      X03.NA <NA> <NA> <NA>
      X04.NA <NA> <NA> <NA>
      symbol
      <factor>
X01.ENSEG00000117616 LOC101928189;RSRP1
X01.ENSEG00000187010 RHCE;RHD
X02.ENSEG00000183726 TMEM50A
X02.ENSEG00000188672 RHCE
      X03.NA <NA>
      X04.NA <NA>

```

seqinfo: 12 sequences from an unspecified genome; no seqlengths

```

> if(interactive()){## annotate the peaks with UCSC annotation
+   library(GenomicFeatures)
+   library(TxDb.Hsapiens.UCSC.hg38.knownGene)
+   ucsc.hg38.knownGene <- genes(TxDb.Hsapiens.UCSC.hg38.knownGene)
+   macs.anno <- annotatePeakInBatch(macsOutput,
+                                 AnnotationData=ucsc.hg38.knownGene,
+                                 output="overlapping", maxgap=5000L)
+   macs.anno <- addGeneIDs(annotatedPeak=macs.anno,
+                           orgAnn="org.Hs.eg.db",
+                           feature_id_type="entrez_id",
+                           IDs2Add="symbol")
+   head(macs.anno)
+ }

```

3 Examples of using ChIPpeakAnno

3.1 Task 1: Find the nearest feature such as gene and the distance to the feature such as the transcription start site (TSS) of the nearest gene

We have a list of peaks identified from ChIP-seq or ChIP-chip experiments and we would like to retrieve the nearest gene and distance to the corresponding gene transcription start site. We have retrieved all the genomic locations of the genes for human genome as TSS.human.NCBI36 data package for repeated use with function `getAnnotation`, now we just pass the annotation to the `annotatePeakInBatch` function.

```
> library(ChIPpeakAnno)
> data(myPeakList)
> data(TSS.human.NCBI36)
> annotatedPeak <- annotatePeakInBatch(myPeakList[1:6,],
+                                   AnnotationData=TSS.human.NCBI36)
> annotatedPeak
```

GRanges object with 6 ranges and 9 metadata columns:

	seqnames	ranges	strand	peak
	<Rle>	<IRanges>	<Rle>	<character>
X1_93_556427.ENSNG00000212875	chr1	[556660, 556760]	*	X1_93_556427
X1_41_559455.ENSNG00000212678	chr1	[559774, 559874]	*	X1_41_559455
X1_12_703729.ENSNG00000197049	chr1	[703885, 703985]	*	X1_12_703729
X1_20_925025.ENSNG00000188290	chr1	[926058, 926158]	*	X1_20_925025
X1_11_1041174.ENSNG00000131591	chr1	[1041646, 1041746]	*	X1_11_1041174
X1_14_1269014.ENSNG00000107404	chr1	[1270239, 1270339]	*	X1_14_1269014

	feature	start_position	end_position
	<character>	<integer>	<integer>
X1_93_556427.ENSNG00000212875	ENSNG00000212875	556318	557859
X1_41_559455.ENSNG00000212678	ENSNG00000212678	559620	560165
X1_12_703729.ENSNG00000197049	ENSNG00000197049	711184	712376
X1_20_925025.ENSNG00000188290	ENSNG00000188290	924209	925333
X1_11_1041174.ENSNG00000131591	ENSNG00000131591	1007062	1041341
X1_14_1269014.ENSNG00000107404	ENSNG00000107404	1260523	1274623

	feature_strand	insideFeature	distancetoFeature
	<character>	<factor>	<numeric>
X1_93_556427.ENSNG00000212875	+	inside	342
X1_41_559455.ENSNG00000212678	+	inside	154
X1_12_703729.ENSNG00000197049	+	upstream	-7299
X1_20_925025.ENSNG00000188290	-	upstream	-725
X1_11_1041174.ENSNG00000131591	-	upstream	-305
X1_14_1269014.ENSNG00000107404	-	inside	4384

	shortestDistance	fromOverlappingOrNearest
	<integer>	<character>
X1_93_556427.ENSNG00000212875	342	NearestLocation
X1_41_559455.ENSNG00000212678	154	NearestLocation
X1_12_703729.ENSNG00000197049	7199	NearestLocation
X1_20_925025.ENSNG00000188290	725	NearestLocation
X1_11_1041174.ENSNG00000131591	305	NearestLocation
X1_14_1269014.ENSNG00000107404	4284	NearestLocation

seqinfo: 24 sequences from an unspecified genome; no seqlengths

To annotate the peaks with other genomic feature, you will need to call function `getAnnotation` with `featureType`, e.g., "Exon" for finding the nearest exon, and "miRNA" for finding the nearest miRNA, "5utr" or "3utr" for finding the overlapping 5 prime UTR or 3 prime UTR. Please refer to `getAnnotation` function for more details.

We have presented the examples using human genome as annotation source. To annotate your data with other species, you will need to pass to the function `getAnnotation` the appropriate dataset for example, `drerio_gene_ensembl` for zebrafish genome, `mmusculus_gene_ensembl` for mouse genome and `rnorvegicus_gene_ensembl` for rat genome.

For a list of available biomaRt and dataset, please refer to the [biomaRt](#) package documentation (Durinck S. et al., 2005). For fast access, in addition to `TSS.human.NCBI36`, `TSS.human.GRCh37`, `TSS.human.GRCh38`, `TSS.mouse.NCBIM37`, `TSS.mouse.GRCm38`, `TSS.rat.RGSC3.4`, `TSS.rat.Rnor_5.0`, `TSS.zebrafish.Zv8`, and `TSS.zebrafish.Zv9` are included as annotation data packages.

You could also pass your own annotation data into the function `annotatePeakInBatch`. For example, if you have a list of transcription factor binding sites from literature and are interested in obtaining the nearest binding site of the transcription factor and distance to it for the list of peaks.

```
> myPeak1 <- GRanges(seqnames=c("1", "2", "3", "4", "5", "6",
+                               "2", "6", "6", "6", "6", "5"),
+                   ranges=IRanges(start=c(967654, 2010897, 2496704, 3075869,
+                                         3123260, 3857501, 201089, 1543200,
+                                         1557200, 1563000, 1569800, 167889600),
+                                   end= c(967754, 2010997, 2496804, 3075969,
+                                         3123360, 3857601, 201089, 1555199,
+                                         1560599, 1565199, 1573799, 167893599),
+                                   names=paste("Site", 1:12, sep="")))
> TFbindingSites <- GRanges(seqnames=c("1", "2", "3", "4", "5", "6", "1", "2", "3",
+                                       "4", "5", "6", "6", "6", "6", "6", "5"),
+                           ranges=IRanges(start=c(967659, 2010898, 2496700,
+                                                   3075866, 3123260, 3857500,
+                                                   96765, 201089, 249670, 307586,
+                                                   312326, 385750, 1549800,
+                                                   1554400, 1565000, 1569400,
+                                                   167888600),
+                                           end=c(967869, 2011108, 2496920,
+                                                  3076166,3123470, 3857780,
+                                                  96985, 201299, 249890, 307796,
+                                                  312586, 385960, 1550599, 1560799,
+                                                  1565399, 1571199, 167888999),
+                                           names=paste("t", 1:17, sep="")),
+                           strand=c("+", "+", "+", "+", "+", "-", "-", "-",
+                                     "-", "-", "-", "+", "+", "+", "+", "+"))
> annotatedPeak2 <- annotatePeakInBatch(myPeak1, AnnotationData=TFbindingSites)
> annotatedPeak2
```

GRanges object with 12 ranges and 9 metadata columns:

	seqnames	ranges	strand	peak	feature
	<Rle>	<IRanges>	<Rle>	<character>	<character>
Site1.t1	chr1	[967654, 967754]	*	Site1	t1
Site2.t2	chr2	[2010897, 2010997]	*	Site2	t2
Site3.t3	chr3	[2496704, 2496804]	*	Site3	t3
Site4.t4	chr4	[3075869, 3075969]	*	Site4	t4
Site5.t5	chr5	[3123260, 3123360]	*	Site5	t5
...
Site8.t14	chr6	[1543200, 1555199]	*	Site8	t14
Site9.t14	chr6	[1557200, 1560599]	*	Site9	t14
Site10.t15	chr6	[1563000, 1565199]	*	Site10	t15
Site11.t16	chr6	[1569800, 1573799]	*	Site11	t16
Site12.t17	chr5	[167889600, 167893599]	*	Site12	t17
	start_position	end_position	feature_strand	insideFeature	distancetoFeature
	<integer>	<integer>	<character>	<factor>	<numeric>
Site1.t1	967659	967869	+	overlapStart	-5

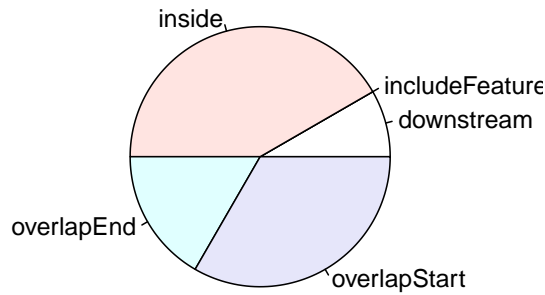


Figure 1: Pie chart of peak distribution among features.

Site2.t2	2010898	2011108	+	overlapStart	-1
Site3.t3	2496700	2496920	+	inside	4
Site4.t4	3075866	3076166	+	inside	3
Site5.t5	3123260	3123470	+	inside	0
...
Site8.t14	1554400	1560799	+	overlapStart	-11200
Site9.t14	1554400	1560799	+	inside	2800
Site10.t15	1565000	1565399	+	overlapStart	-2000
Site11.t16	1569400	1571199	+	overlapEnd	400
Site12.t17	167888600	167888999	+	downstream	1000

	shortestDistance	fromOverlappingOrNearest
	<integer>	<character>
Site1.t1	5	NearestLocation
Site2.t2	1	NearestLocation
Site3.t3	4	NearestLocation
Site4.t4	3	NearestLocation
Site5.t5	0	NearestLocation
...
Site8.t14	799	NearestLocation
Site9.t14	200	NearestLocation
Site10.t15	199	NearestLocation
Site11.t16	400	NearestLocation
Site12.t17	601	NearestLocation

seqinfo: 6 sequences from an unspecified genome; no seqlengths

```
> pie(table(as.data.frame(annotatedPeak2)$insideFeature))
```

Both BED format and GFF format are common file format that provides a flexible way to define the peaks and annotations as the data lines. Therefore, conversion functions toGRanges were implemented for converting these data format to GRanges before calling `annotatePeakInBatch`

Once you annotated the peak list, you can plot the distance to nearest feature such as TSS.

3.2 Task 2: Obtain overlapping peaks for potential transcription factor complex and determine the significance of the overlapping and generate Venn Diagram

Here is an example of obtaining overlapping peaks with maximum gap 1kb for two peak ranges.

```
> peaks1 <- GRanges(seqnames=c("1", "2", "3", "4", "5", "6",
+                               "2", "6", "6", "6", "6", "5"),
+                   ranges=IRanges(start=c(967654, 2010897, 2496704, 3075869,
+                                         3123260, 3857501, 201089, 1543200,
+                                         1557200, 1563000, 1569800, 167889600),
+                                   end= c(967754, 2010997, 2496804, 3075969,
+                                         3123360, 3857601, 201089, 1555199,
+                                         1560599, 1565199, 1573799, 167893599),
+                                   names=paste("Site", 1:12, sep="")),
+                   strand="+")
> peaks2 <- GRanges(seqnames=c("1", "2", "3", "4", "5", "6", "1", "2", "3",
+                               "4", "5", "6", "6", "6", "6", "6", "5"),
+                   ranges=IRanges(start=c(967659, 2010898, 2496700,
+                                         3075866, 3123260, 3857500,
+                                         96765, 201089, 249670, 307586,
+                                         312326, 385750, 1549800,
+                                         1554400, 1565000, 1569400,
+                                         167888600),
+                                   end=c(967869, 2011108, 2496920,
+                                         3076166,3123470, 3857780,
+                                         96985, 201299, 249890, 307796,
+                                         312586, 385960, 1550599, 1560799,
+                                         1565399, 1571199, 167888999),
+                                   names=paste("t", 1:17, sep="")),
+                   strand=c("+", "+", "+", "+", "+", "+", "-", "-", "-",
+                               "-", "-", "-", "+", "+", "+", "+", "+"))
> ol <- findOverlapsOfPeaks(peaks1, peaks2, maxgap=1000)
> peaklist <- ol$peaklist
```

Here is a list of overlapping peaks with maximum gap 1kb and a pie graph describing the distribution of relative position of peaks1 to peaks2 for overlapping peaks.

```
> overlappingPeaks <- ol$overlappingPeaks
> overlappingPeaks

$`peaks1///peaks2`
      peaks1 seqnames      start      end width strand
peaks1__Site1_peaks2__t1 peaks1__Site1      1  967654  967754  101  +
peaks1__Site7_peaks2__t8 peaks1__Site7      2  201089  201089   1  +
peaks1__Site2_peaks2__t2 peaks1__Site2      2 2010897 2010997  101  +
peaks1__Site3_peaks2__t3 peaks1__Site3      3 2496704 2496804  101  +
peaks1__Site4_peaks2__t4 peaks1__Site4      4 3075869 3075969  101  +
peaks1__Site5_peaks2__t5 peaks1__Site5      5 3123260 3123360  101  +
peaks1__Site12_peaks2__t17 peaks1__Site12     5 167889600 167893599 4000  +
peaks1__Site8_peaks2__t13 peaks1__Site8      6 1543200 1555199 12000  +
peaks1__Site8_peaks2__t14 peaks1__Site8      6 1543200 1555199 12000  +
peaks1__Site9_peaks2__t14 peaks1__Site9      6 1557200 1560599 3400  +
peaks1__Site10_peaks2__t15 peaks1__Site10     6 1563000 1565199 2200  +
peaks1__Site11_peaks2__t16 peaks1__Site11     6 1569800 1573799 4000  +
peaks1__Site6_peaks2__t6 peaks1__Site6      6 3857501 3857601  101  +
      peaks2 seqnames      start      end width strand
peaks1__Site1_peaks2__t1 peaks2__t1      1  967659  967869  211  +
peaks1__Site7_peaks2__t8 peaks2__t8      2  201089  201299  211  -
peaks1__Site2_peaks2__t2 peaks2__t2      2 2010898 2011108  211  +
peaks1__Site3_peaks2__t3 peaks2__t3      3 2496700 2496920  221  +
peaks1__Site4_peaks2__t4 peaks2__t4      4 3075866 3076166  301  +
```

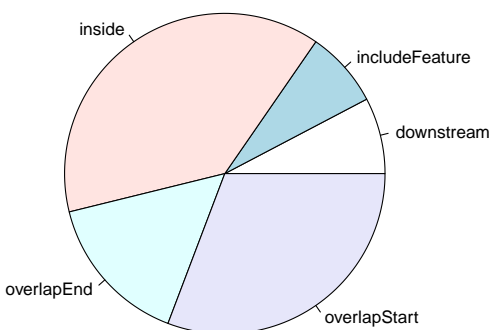


Figure 2: Pie chart of common peaks among features.

```

peaks1__Site5_peaks2__t5 peaks2__t5 5 3123260 3123470 211 +
peaks1__Site12_peaks2__t17 peaks2__t17 5 167888600 167888999 400 +
peaks1__Site8_peaks2__t13 peaks2__t13 6 1549800 1550599 800 +
peaks1__Site8_peaks2__t14 peaks2__t14 6 1554400 1560799 6400 +
peaks1__Site9_peaks2__t14 peaks2__t14 6 1554400 1560799 6400 +
peaks1__Site10_peaks2__t15 peaks2__t15 6 1565000 1565399 400 +
peaks1__Site11_peaks2__t16 peaks2__t16 6 1569400 1571199 1800 +
peaks1__Site6_peaks2__t6 peaks2__t6 6 3857500 3857780 281 +

```

```

overlapFeature shortestDistance
peaks1__Site1_peaks2__t1 overlapStart 5
peaks1__Site7_peaks2__t8 overlapEnd 0
peaks1__Site2_peaks2__t2 overlapStart 1
peaks1__Site3_peaks2__t3 inside 4
peaks1__Site4_peaks2__t4 inside 3
peaks1__Site5_peaks2__t5 inside 0
peaks1__Site12_peaks2__t17 downstream 601
peaks1__Site8_peaks2__t13 includeFeature 4600
peaks1__Site8_peaks2__t14 overlapStart 799
peaks1__Site9_peaks2__t14 inside 200
peaks1__Site10_peaks2__t15 overlapStart 199
peaks1__Site11_peaks2__t16 overlapEnd 400
peaks1__Site6_peaks2__t6 inside 1

```

```
> pie(table(overlappingPeaks[["peaks1///peaks2"]]$overlapFeature))
```

Here is the merged overlapping peaks, which can be used to obtain overlapping peaks with another TF binding sites from a protein complex.

```
> peaklist[["peaks1///peaks2"]]
```

GRanges object with 11 ranges and 1 metadata column:

```

seqnames      ranges strand |
  <Rle>        <IRanges> <Rle> |
[1]          1 [ 967654, 967869] + |
[2]          2 [ 201089, 201299] * |

```



```

[3]      2      [2010897, 2011108]      + |
[4]      3      [2496700, 2496920]      + |
[5]      4      [3075866, 3076166]      + |
...      ...      ...      ...      ...
[7]      5 [167888600, 167893599]      + |
[8]      6 [ 1543200, 1560799]      + |
[9]      6 [ 1563000, 1565399]      + |
[10]     6 [ 1569400, 1573799]      + |
[11]     6 [ 3857500, 3857780]      + |

```

```

                                peakNames
                                <CharacterList>
[1] peaks1__Site1,peaks2__t1
[2] peaks1__Site7,peaks2__t8
[3] peaks1__Site2,peaks2__t2
[4] peaks2__t3,peaks1__Site3
[5] peaks2__t4,peaks1__Site4
...      ...
[7] peaks2__t17,peaks1__Site12
[8] peaks1__Site8,peaks2__t13,peaks2__t14,...
[9] peaks1__Site10,peaks2__t15
[10] peaks2__t16,peaks1__Site11
[11] peaks2__t6,peaks1__Site6
-----

```

seqinfo: 6 sequences from an unspecified genome; no seqlengths

Here is the peaks in peaks1 that not overlaps with peaks in peaks2

```
> peaklist[["peaks1"]]
```

NULL

Here is the peaks in peaks2 that not overlap with peaks in peaks1

```
> peaklist[["peaks2"]]
```

GRanges object with 5 ranges and 1 metadata column:

```

seqnames      ranges strand |      peakNames
  <Rle>      <IRanges> <Rle> | <CharacterList>
[1]      1 [ 96765, 96985] - |      peaks2__t7
[2]      3 [249670, 249890] - |      peaks2__t9
[3]      4 [307586, 307796] - |      peaks2__t10
[4]      5 [312326, 312586] - |      peaks2__t11
[5]      6 [385750, 385960] - |      peaks2__t12
-----

```

seqinfo: 6 sequences from an unspecified genome; no seqlengths

Venn Diagram can be generated by the following function call using the results of `findOverlapsOfPeaks` as an input (Figure 3). P-values indicate whether the extent of overlapping is significant.

```
> makeVennDiagram(ol, totalTest=1e+2)
```

```

$p.p.value
  peaks1 peaks2      pval
[1,]     1     1 5.890971e-12

```

```

$vennCounts
  peaks1 peaks2 Counts
[1,]     0     0     83
[2,]     0     1     5
[3,]     1     0     0
[4,]     1     1    12

```

```

attr(,"class")
[1] "VennCounts"

```

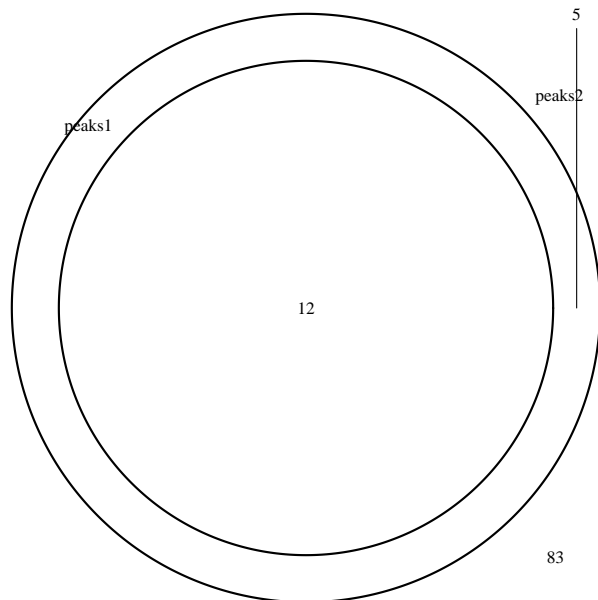


Figure 3: venn diagram of overlaps

Users can also try other tools to draw vennDiagrams such as *Ven venerable*.

```
> # install.packages("Ven venerable", repos="http://R-Forge.R-project.org", type="source")
> # library(Ven venerable)
> # venn_cnt2venn <- function(venn_cnt){
> #     n <- which(colnames(venn_cnt)=="Counts") - 1
> #     SetNames=colnames(venn_cnt)[1:n]
> #     Weight=venn_cnt[, "Counts"]
> #     names(Weight) <- apply(venn_cnt[,1:n], 1, paste, collapse="")
> #     Venn(SetNames=SetNames, Weight=Weight)
> # }
> #
> # v <- venn_cnt2venn(ol$venn_cnt)
> # plot(v)
```

The `findOverlapsOfPeaks` function can be called to obtain overlaps upto 5 peak lists for example, the overlap peaks in peaks1, peaks2 and peaks3 (Figure 4).

```
> peaks3 <- GRanges(seqnames=c("1", "2", "3", "4", "5",
+                               "6", "1", "2", "3", "4"),
+                   ranges=IRanges(start=c(967859, 2010868, 2496500, 3075966,
+                                           3123460, 3851500, 96865, 201189,
+                                           249600, 307386),
+                                   end= c(967969, 2011908, 2496720, 3076166,
+                                           3123470, 3857680, 96985, 201299,
+                                           249890, 307796),
+                                   names=paste("p", 1:10, sep="")),
+                   strand=c("+", "+", "+", "+", "+",
+                              "+", "-", "-", "-", "-"))
> ol <- findOverlapsOfPeaks(peaks1, peaks2, peaks3, maxgap=1000, connectedPeaks="min")
> makeVennDiagram(ol, totalTest=1e+2)
```

```
$p.value
  peaks1 peaks2 peaks3      pval
[1,]    0      1      1 1.123492e-09
```

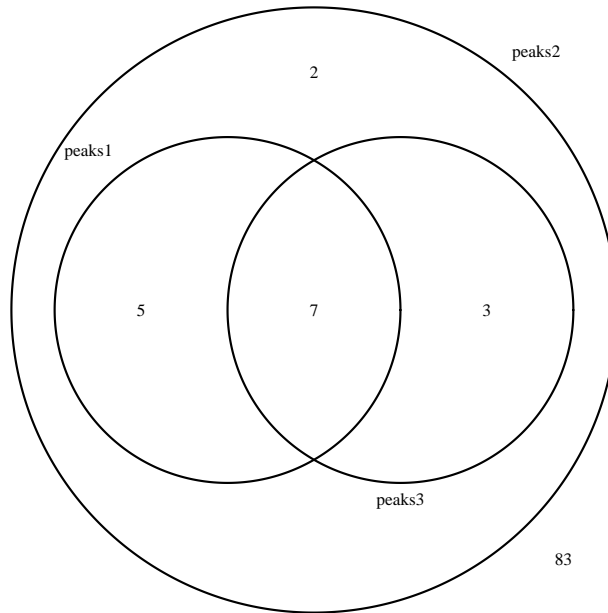


Figure 4: venn diagram of overlaps for three input peak lists

```
[2,] 1 0 1 5.131347e-06
[3,] 1 1 0 5.890971e-12
```

```
$vennCounts
  peaks1 peaks2 peaks3 Counts
[1,]    0     0     0     83
[2,]    0     0     1     0
[3,]    0     1     0     2
[4,]    0     1     1     3
[5,]    1     0     0     0
[6,]    1     0     1     0
[7,]    1     1     0     5
[8,]    1     1     1     7
attr(,"class")
[1] "VennCounts"
```

Venn Diagram can also be generated by the following function call with p-value that indicates whether the extent of overlapping is significant (Figure 5,6). Note, the maxgap is changed to 0.

```
> makeVennDiagram(list(peaks1, peaks2), NameOfPeaks=c("TF1", "TF2"),
+                 maxgap=0, minoverlap =1, totalTest=100)
```

```
$p.value
  TF1 TF2      pval
[1,] 1  1 9.837922e-10
```

```
$vennCounts
  TF1 TF2 Counts
[1,] 0 0 82
[2,] 0 1 6
[3,] 1 0 1
[4,] 1 1 11
attr(,"class")
[1] "VennCounts"
```

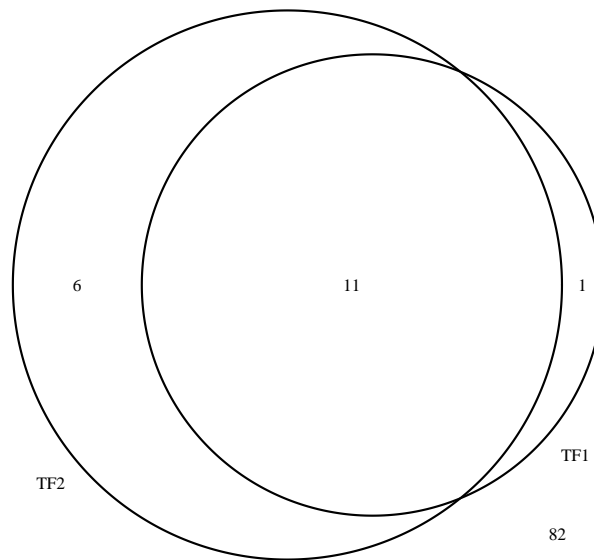


Figure 5: Venn diagram to depict the overlaps between two peak lists

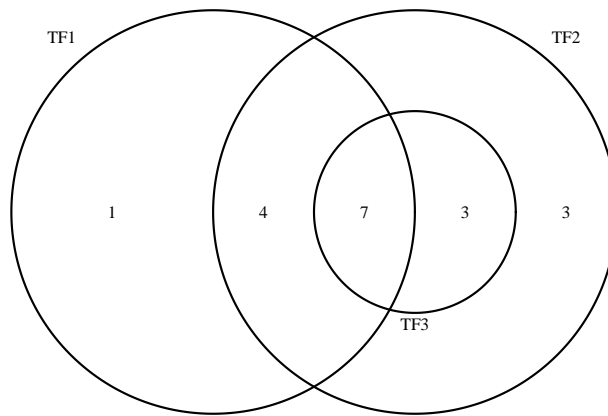
```
> makeVennDiagram(list(peaks1, peaks2, peaks3),
+                 NameOfPeaks=c("TF1", "TF2", "TF3"),
+                 maxgap=0, minoverlap =1, totalTest=100)
```

```
$p.value
      TF1 TF2 TF3      pval
[1,]   0   1   1 1.123492e-09
[2,]   1   0   1 5.131347e-06
[3,]   1   1   0 9.837922e-10
```

```
$vennCounts
      TF1 TF2 TF3 Counts
[1,]   0   0   0      82
[2,]   0   0   1       0
[3,]   0   1   0       3
[4,]   0   1   1       3
[5,]   1   0   0       1
[6,]   1   0   1       0
[7,]   1   1   0       4
[8,]   1   1   1       7
attr(,"class")
[1] "VennCounts"
```

3.3 Task 3: Obtain sequences surrounding the peaks for PCR validation or motif discovery

Here is an example of obtaining sequences surrounding the peak intervals including 20 bp upstream and downstream sequence.



82

Figure 6: venn diagram of overlaps for three input peaklists directly

```
> peaks <- GRanges(seqnames=c("NC_008253", "NC_010468"),
+                 ranges=IRanges(start=c(100, 500),
+                               end=c(300, 600),
+                               names=c("peak1", "peak2")))
> library(BSgenome.Ecoli.NCBI.20080805)
> peaksWithSequences <- getAllPeakSequence(peaks, upstream=20,
+                                          downstream=20, genome=Ecoli)
```

You can easily convert the obtained sequences into fasta format for motif discovery by calling the function `write2FASTA`.

```
> write2FASTA(peaksWithSequences, "test.fa")
```

3.4 Task 4: Obtain enriched gene ontology (GO) terms or KEGG terms near the peaks

Once you have obtained the annotated peak data from the example above, you can also use the function `getEnriched` to obtain a list of enriched gene ontology (GO) terms via [GOstats](#). The ontology could also be set as KEGG or reactome.

Once you have obtained the annotated peak data from the example above, you can also use the function `getEnrichedGO` to obtain a list of enriched gene ontology (GO) terms using hypergeometric test.

```
library(org.Hs.eg.db)
```

```
enrichedGO = getEnrichedGO (annotatedPeak, orgAnn = "org.Hs.eg.db", maxP = 0.01,
multiAdj = TRUE, minGOterm = 10, multiAdjMethod = "BH" )
```

```
> library(org.Hs.eg.db)
> over <- getEnrichedGO(annotatedPeak, orgAnn="org.Hs.eg.db",
+ maxP=0.01, multiAdj=FALSE, minGOterm=10, multiAdjMethod="")
> head(over[["bp"]])
```

```
      go.id          go.term
1 GO:0001736  establishment of planar polarity
2 GO:0001840      neural plate development
3 GO:0001941  postsynaptic membrane organization
4 GO:0001964      startle response
5 GO:0007164  establishment of tissue polarity
6 GO:0031122  cytoplasmic microtubule organization
```

```
1
2 The process whose specific outcome is the progression of the neural plate over time, from its formation to the mature structure
3
4
5
6
```

	Ontology	count.InDataset	count.InGenome	pvalue	totaltermInDataset
1	BP	1	28	0.008619994	405
2	BP	1	11	0.003395307	405
3	BP	1	22	0.006779114	405
4	BP	1	23	0.007086164	405
5	BP	1	28	0.008619994	405
6	BP	1	32	0.009845356	405

	totaltermInGenome	EntrezID
1	1310084	1855
2	1310084	1855
3	1310084	1855
4	1310084	1855
5	1310084	1855
6	1310084	1855

```
> head(over[["cc"]])
```

```
      go.id          go.term
1 GO:0016328  lateral plasma membrane
```

```
1 The portion of the plasma membrane at the lateral side of the cell. In epithelial cells, lateral plasma membranes are on the surface
```

	Ontology	count.InDataset	count.InGenome	pvalue	totaltermInDataset
1	CC	1	48	0.008016845	61

	totaltermInGenome	EntrezID
1	363819	1855

```
> head(over[["mf"]])
```

```
      go.id          go.term
1 GO:0005109  frizzled binding
2 GO:0017048  Rho GTPase binding
3 GO:0048365  Rac GTPase binding
```

```
1
2 Interacting selectively and non-covalently with Rho protein, any member of the Rho subfamily of the Ras superfamily of monomeric GTPases
3
```

	Ontology	count.InDataset	count.InGenome	pvalue	totaltermInDataset
1	MF	1	37	0.003861301	24
2	MF	1	71	0.007396923	24
3	MF	1	32	0.003340340	24

	totaltermInGenome	EntrezID
1	229560	1855
2	229560	1855

Please note that org.Hs.eg.db is the GO gene mapping for Human, for other organisms, please refer to <http://www.bioconductor.org/packages/release/data/annotation/> for additional org.xx.eg.db packages. Or you can try egOrgMap to get the annotation database.

```
> egOrgMap("Mus musculus")
[1] "org.Mm.eg.db"
> egOrgMap("Homo sapiens")
[1] "org.Hs.eg.db"
```

3.5 Task 5: Find peaks with bi-directional promoters

Here is an example to find peaks with bi-directional promoters and output percent of peaks near bi-directional promoters.

```
> data(myPeakList)
> data(TSS.human.NCBI36)
> annotatedBDP <- peaksNearBDP(myPeakList[1:10,],
+                               AnnotationData=TSS.human.NCBI36,
+                               MaxDistance=5000,
+                               PeakLocForDistance="middle",
+                               FeatureLocForDistance="TSS")
> annotatedBDP$peaksWithBDP
```

GRanges object with 6 ranges and 9 metadata columns:

	seqnames	ranges	strand	peak
	<Rle>	<IRanges>	<Rle>	<character>
X1_14_1300250.ENSG00000218550	chr1	[1300503, 1300603]	*	X1_14_1300250
X1_14_1300250.ENSG00000175756	chr1	[1300503, 1300603]	*	X1_14_1300250
X1_41_559455.ENSG00000212678	chr1	[559774, 559874]	*	X1_41_559455
X1_41_559455.ENSG00000209350	chr1	[559774, 559874]	*	X1_41_559455
X1_93_556427.ENSG00000212875	chr1	[556660, 556760]	*	X1_93_556427
X1_93_556427.ENSG00000209349	chr1	[556660, 556760]	*	X1_93_556427
	feature	start_position	end_position	
	<character>	<integer>	<integer>	
X1_14_1300250.ENSG00000218550	ENSG00000218550	1303908	1304275	
X1_14_1300250.ENSG00000175756	ENSG00000175756	1298974	1300443	
X1_41_559455.ENSG00000212678	ENSG00000212678	559620	560165	
X1_41_559455.ENSG00000209350	ENSG00000209350	557860	557930	
X1_93_556427.ENSG00000212875	ENSG00000212875	556318	557859	
X1_93_556427.ENSG00000209349	ENSG00000209349	556240	556304	
	feature_strand	insideFeature	distancetoFeature	
	<character>	<factor>	<numeric>	
X1_14_1300250.ENSG00000218550	+	upstream	-3355	
X1_14_1300250.ENSG00000175756	-	upstream	-110	
X1_41_559455.ENSG00000212678	+	inside	204	
X1_41_559455.ENSG00000209350	-	upstream	-1894	
X1_93_556427.ENSG00000212875	+	inside	392	
X1_93_556427.ENSG00000209349	-	upstream	-406	
	shortestDistance	fromOverlappingOrNearest		
	<integer>	<character>		
X1_14_1300250.ENSG00000218550	3305	NearestLocation		
X1_14_1300250.ENSG00000175756	60	NearestLocation		
X1_41_559455.ENSG00000212678	154	NearestLocation		
X1_41_559455.ENSG00000209350	1844	NearestLocation		
X1_93_556427.ENSG00000212875	342	NearestLocation		
X1_93_556427.ENSG00000209349	356	NearestLocation		

```

-----
seqinfo: 24 sequences from an unspecified genome; no seqlengths
> c(annotatedBDP$percentPeaksWithBDP,
+   annotatedBDP$n.peaks,
+   annotatedBDP$n.peaksWithBDP)
[1] 0.3 10.0 3.0

```

3.6 Task 6: Output a summary of motif occurrence in the peaks.

Here is an example to search the peaks for the motifs in examplepattern.fa file.

```

> peaks <- GRanges(seqnames=c("NC_008253", "NC_010468"),
+                 ranges=IRanges(start=c(100, 500),
+                                 end=c(300, 600),
+                                 names=c("peak1", "peak2")))
> filepath <- system.file("extdata", "examplePattern.fa", package="ChIPpeakAnno")
> library(BSgenome.Ecoli.NCBI.20080805)
> summarizePatternInPeaks(patternFilePath=filepath, format="fasta", skip=0L,
+                          BSgenomeName=Ecoli, peaks=peaks)

```

	n.peaksWithPattern	n.totalPeaks	Pattern
[1,]	"0"	"2"	"GGNCCK"
[2,]	"1"	"2"	"AACCNM"

3.7 Task 7: Add other IDs to annotated peaks or enrichedGO

Here is an example to add gene symbol to annotated peaks .

```

> data(annotatedPeak)
> library(org.Hs.eg.db)
> addGeneIDs(annotatedPeak[1:6,], orgAnn="org.Hs.eg.db", IDs2Add=c("symbol"))

```

GRanges object with 6 ranges and 9 metadata columns:

seqnames	ranges	strand	peak
<Rle>	<IRanges>	<Rle>	<character>
X1_11_100272487.ENSEG00000202254	1 [100272801, 100272900]	+	1_11_100272487
X1_11_108905539.ENSEG00000186086	1 [108906026, 108906125]	+	1_11_108905539
X1_11_110106925.ENSEG00000065135	1 [110107267, 110107366]	+	1_11_110106925
X1_11_110679983.ENSEG00000197106	1 [110680469, 110680568]	+	1_11_110679983
X1_11_110681677.ENSEG00000197106	1 [110682125, 110682224]	+	1_11_110681677
X1_11_110756560.ENSEG00000116396	1 [110756823, 110756922]	+	1_11_110756560

feature	start_position	end_position
<character>	<numeric>	<numeric>
X1_11_100272487.ENSEG00000202254	100257218	100257309
X1_11_108905539.ENSEG00000186086	108918435	109013624
X1_11_110106925.ENSEG00000065135	110091233	110136975
X1_11_110679983.ENSEG00000197106	110693108	110744824
X1_11_110681677.ENSEG00000197106	110693108	110744824
X1_11_110756560.ENSEG00000116396	110753965	110776666

insideFeature	distancetoFeature	shortestDistance
<character>	<numeric>	<numeric>
X1_11_100272487.ENSEG00000202254	downstream	15582
X1_11_108905539.ENSEG00000186086	upstream	-12410
X1_11_110106925.ENSEG00000065135	inside	16033
X1_11_110679983.ENSEG00000197106	upstream	-12640
X1_11_110681677.ENSEG00000197106	upstream	-10984
X1_11_110756560.ENSEG00000116396	inside	2857


```

                                fromOverlappingOrNearest  symbol
                                <character> <factor>
X1_11_100272487.ENSEG00000202254      NearestStart    <NA>
X1_11_108905539.ENSEG00000186086      NearestStart    NBPF6
X1_11_110106925.ENSEG00000065135      NearestStart    GNAI3
X1_11_110679983.ENSEG00000197106      NearestStart    SLC6A17
X1_11_110681677.ENSEG00000197106      NearestStart    SLC6A17
X1_11_110756560.ENSEG00000116396      NearestStart    KCNC4
-----
seqinfo: 24 sequences from an unspecified genome; no seqlengths
> addGeneIDs(annotatedPeak$feature[1:6], orgAnn="org.Hs.eg.db", IDs2Add=c("symbol"))

ensembl_gene_id  symbol
1 ENSG00000065135  GNAI3
2 ENSG00000116396  KCNC4
3 ENSG00000197106  SLC6A17
4 ENSG00000186086  NBPF6
5 ENSG00000202254  <NA>

```

3.8 Task 8: annotate ChIP results from BED or GFF files or MACS output xls file

Here is an example to annotate peaks in BED file format and GFF file format.

```

> bed <- system.file("extdata", "MACS_output.bed", package="ChIPpeakAnno")
> gr1 <- toGRanges(bed, format="BED", header=FALSE)
> ## one can also try import from rtracklayer
> library(rtracklayer)
> gr1.import <- import(bed, format="BED")
> identical(start(gr1), start(gr1.import))

[1] TRUE

> gr1[1:2]

GRanges object with 2 ranges and 1 metadata column:
      seqnames      ranges strand |      score
      <Rle>        <IRanges> <Rle> | <numeric>
MACS_peak_1      chr1 [28341, 29610]   * |    160.81
MACS_peak_2      chr1 [90821, 91234]   * |    133.12
-----
seqinfo: 1 sequence from an unspecified genome; no seqlengths

> gr1.import[1:2] #note the name slot is different from gr1

GRanges object with 2 ranges and 2 metadata columns:
      seqnames      ranges strand |      name      score
      <Rle>        <IRanges> <Rle> | <character> <numeric>
[1]      chr1 [28341, 29610]   * | MACS_peak_1    160.81
[2]      chr1 [90821, 91234]   * | MACS_peak_2    133.12
-----
seqinfo: 1 sequence from an unspecified genome; no seqlengths

> gff <- system.file("extdata", "GFF_peaks.gff", package="ChIPpeakAnno")
> gr2 <- toGRanges(gff, format="GFF", header=FALSE, skip=3)
> ol <- findOverlapsOfPeaks(gr1, gr2)
> makeVennDiagram(ol)

$p.value
      gr1 gr2 pval
[1,]  1  1  0

```

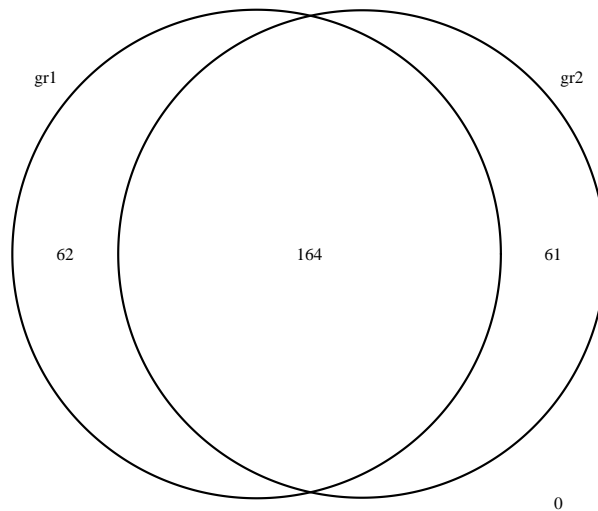


Figure 7: venn diagram of overlaps for duplicated experiments

```
$vennCounts
      gr1 gr2 Counts
[1,]   0   0     0
[2,]   0   1     61
[3,]   1   0     62
[4,]   1   1    164
attr(,"class")
[1] "VennCounts"
```

```
> pie(table(ol$overlappingPeaks[["gr1//gr2"]]$overlapFeature))
```

Find all features within 5kb away from the overlapping peaks using `annotatePeakInBatch`.

```
> data(TSS.human.GRCh37)
> overlaps <- ol$peaklist[["gr1//gr2"]]
> overlaps.anno <- annotatePeakInBatch(overlaps, AnnotationData=TSS.human.GRCh37,
+                                     output="overlapping", maxgap=5000L)
> overlaps.anno <- addGeneIDs(overlaps.anno, "org.Hs.eg.db", "symbol")
> head(overlaps.anno)
```

GRanges object with 6 ranges and 11 metadata columns:

	seqnames	ranges	strand	
	<Rle>	<IRanges>	<Rle>	
X001.ENSG00000228327	chr1	[713791, 715578]	*	
X001.ENSG00000237491	chr1	[713791, 715578]	*	
X001.ENSG00000242937	chr1	[713791, 715578]	*	
X002.ENSG00000237491	chr1	[724851, 727191]	*	
X002.ENSG00000242937	chr1	[724851, 727191]	*	
X002.ENSG00000197049	chr1	[724851, 727191]	*	

	peakNames	peak
	<CharacterList>	<character>
X001.ENSG00000228327	gr1__MACS_peak_13,gr2__region_0,gr2__region_1	001
X001.ENSG00000237491	gr1__MACS_peak_13,gr2__region_0,gr2__region_1	001
X001.ENSG00000242937	gr1__MACS_peak_13,gr2__region_0,gr2__region_1	001

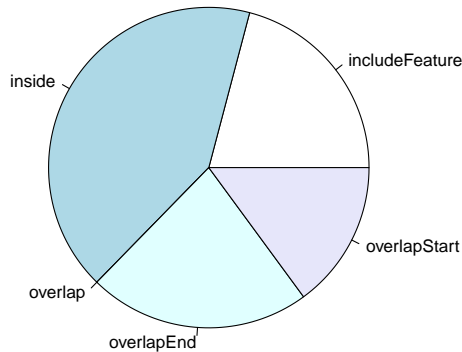


Figure 8: Pie chart of common peaks among features

```

X002.ENSEG00000237491      gr2__region_2,gr1__MACS_peak_14      002
X002.ENSEG00000242937      gr2__region_2,gr1__MACS_peak_14      002
X002.ENSEG00000197049      gr2__region_2,gr1__MACS_peak_14      002
      feature start_position end_position feature_strand
      <character>      <integer>      <integer>      <character>
X001.ENSEG00000228327      ENSG00000228327      700238      714006      -
X001.ENSEG00000237491      ENSG00000237491      714163      740255      +
X001.ENSEG00000242937      ENSG00000242937      717326      720070      +
X002.ENSEG00000237491      ENSG00000237491      714163      740255      +
X002.ENSEG00000242937      ENSG00000242937      717326      720070      +
X002.ENSEG00000197049      ENSG00000197049      721321      722513      +
      insideFeature distancetoFeature shortestDistance
      <factor>      <numeric>      <integer>
X001.ENSEG00000228327      overlapStart      215      215
X001.ENSEG00000237491      overlapStart      -372      372
X001.ENSEG00000242937      upstream      -3535      1748
X002.ENSEG00000237491      inside      10688      10688
X002.ENSEG00000242937      downstream      7525      4781
X002.ENSEG00000197049      downstream      3530      2338
      fromOverlappingOrNearest      symbol
      <character>      <factor>
X001.ENSEG00000228327      Overlapping      LOC100288069;LOC101929540
X001.ENSEG00000237491      Overlapping      LOC100287934
X001.ENSEG00000242937      Overlapping      <NA>
X002.ENSEG00000237491      Overlapping      LOC100287934
X002.ENSEG00000242937      Overlapping      <NA>
X002.ENSEG00000197049      Overlapping      <NA>

```

seqinfo: 1 sequence from an unspecified genome; no seqlengths

Plot the distribution of aggregated peak scores or peak numbers around transcript start sites (Figure 9).

```

> gr1.copy <- gr1
> gr1.copy$score <- 1

```

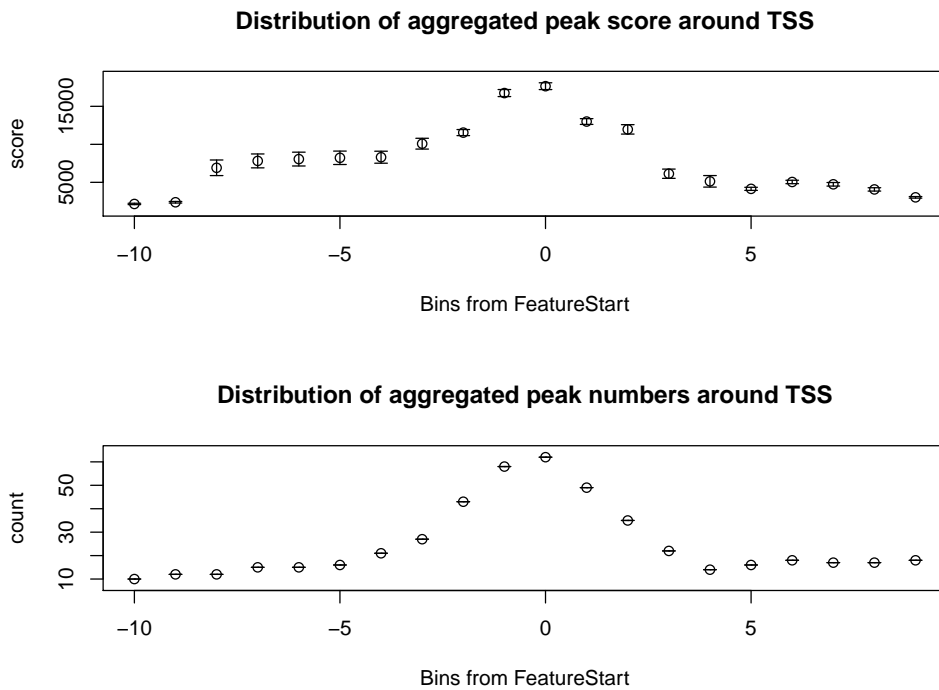


Figure 9: Distribution of aggregated peak scores or peak numbers around transcript start sites.

```
> binOverFeature(gr1, gr1.copy, annotationData=TSS.human.GRCh37,
+               radius=5000, nbins=10, FUN=c(sum, length),
+               ylab=c("score", "count"),
+               main=c("Distribution of aggregated peak score around TSS",
+                     "Distribution of aggregated peak numbers around TSS"))
```

Summarize peak distribution over exon, intron, enhancer, proximal promoter, 5 prime UTR and 3 prime UTR in peak centric and nucleotide centric view using function `assignChromosomeRegion`(Figure 10). Setting `nucleotideLevel = TRUE` will give a nucleotide level distribution over different features.

```
> if(require(TxDb.Hsapiens.UCSC.hg19.knownGene)){
+   aCR<-assignChromosomeRegion(gr1, nucleotideLevel=FALSE,
+                             precedence=c("Promoters", "immediateDownstream",
+                                           "fiveUTRs", "threeUTRs",
+                                           "Exons", "Introns"),
+                             TxDb=TxDb.Hsapiens.UCSC.hg19.knownGene)
+   barplot(aCR$percentage)
+ }
```

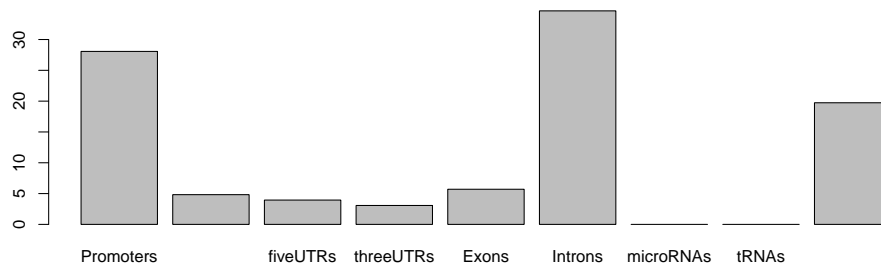


Figure 10: Peak distribution over different genomic features.

4 References

1. Y. Benjamini and Y. Hochberg (1995). Controlling the false discovery rate: a practical and powerful approach to multiple testing. *J. R. Statist. Soc. B.* Vol. 57: 289-300.
2. Y. Benjamini and D. Yekutieli (2001). The control of the false discovery rate in multiple hypothesis testing under dependency. *Annals of Statistics*. Accepted.
3. S. Durinck et al. (2005) BioMart and Bioconductor: a powerful link between biological biomarts and microarray data analysis. *Bioinformatics*, 21, 3439-3440.
4. S. Dudoit, J. P. Shaffer, and J. C. Boldrick (Submitted). Multiple hypothesis testing in microarray experiments.
5. Y. Ge, S. Dudoit, and T. P. Speed. Resampling-based multiple testing for microarray data hypothesis, Technical Report #633 of UCB Stat. <http://www.stat.berkeley.edu/gyc>
6. R. Gentleman et al. (2004) Bioconductor: open software development for computational biology and bioinformatics. *Genome Biol.*, 5:R80
7. Y. Hochberg (1988). A sharper Bonferroni procedure for multiple tests of significance, *Biometrika*. Vol. 75: 800-802.
8. S. Holm (1979). A simple sequentially rejective multiple test procedure. *Scand. J. Statist.* Vol. 6: 65-70.
9. N. L. Johnson, S. Kotz and A. W. Kemp (1992) *Univariate Discrete Distributions*, Second Edition. New York: Wiley
10. G. Robertson et al. (2007) Genome-wide profiles of STAT1 DNA association using chromatin immunoprecipitation and massively parallel sequencing. *Nat Methods*, 4:651-7.
11. Zhu L.J. et al. (2010) ChIPpeakAnno: a Bioconductor package to annotate ChIP-seq and ChIP-chip data. *BMC Bioinformatics* 2010, 11:237doi:10.1186/1471-2105-11-237.
12. Zhu L.J. (2013) Integrative analysis of ChIP-chip and ChIP-seq dataset. *Methods Mol Biol.* 2013;1067:105-24. doi: 10.1007/978-1-62703-607-8_8.

5 Session Info

```
> toLatex(sessionInfo())
```

- R version 3.2.0 (2015-04-16), x86_64-unknown-linux-gnu
- Locale: LC_CTYPE=en_US.UTF-8, LC_NUMERIC=C, LC_TIME=en_US.UTF-8, LC_COLLATE=C, LC_MONETARY=en_US.UTF-8, LC_MESSAGES=en_US.UTF-8, LC_PAPER=en_US.UTF-8, LC_NAME=C, LC_ADDRESS=C, LC_TELEPHONE=C, LC_MEASUREMENT=en_US.UTF-8, LC_IDENTIFICATION=C
- Base packages: base, datasets, grDevices, graphics, grid, methods, parallel, stats, stats4, utils
- Other packages: AnnotationDbi 1.30.1, BSgenome 1.36.0, BSgenome.Ecoli.NCBI.20080805 1.3.1000, Biobase 2.28.0, BiocGenerics 0.14.0, Biostrings 2.36.1, ChIPpeakAnno 3.2.2, DBI 0.3.1, FDb.UCSC.tRNAs 1.0.1, GenomInfoDb 1.4.0, GenomicFeatures 1.20.1, GenomicRanges 1.20.3, IRanges 2.2.1, RSQLite 1.0.0, S4Vectors 0.6.0, TxDb.Hsapiens.UCSC.hg19.knownGene 3.1.2, VennDiagram 1.6.9, XVector 0.8.0, biomaRt 2.24.0, mirbase.db 1.2.0, org.Hs.eg.db 3.1.2, rtracklayer 1.28.2
- Loaded via a namespace (and not attached): BiocInstaller 1.18.2, BiocParallel 1.2.1, BiocStyle 1.6.0, GO.db 3.1.2, GenomicAlignments 1.4.1, MASS 7.3-40, RBGL 1.44.0, RCurl 1.95-4.6, Rsamtools 1.20.2, XML 3.98-1.1, bitops 1.0-6, futile.logger 1.4.1, futile.options 1.0.0, graph 1.46.0, lambda.r 1.1.7, limma 3.24.4, multtest 2.24.0, splines 3.2.0, survival 2.38-1, tools 3.2.0, zlibbioc 1.14.0