# The Gviz User Guide

#### Florian Hahne\*

#### October 14, 2013

# Contents

1	Introduction				
2	2 Basic Features				
3	3 Plotting parameters	8			
4	Track classes	13			
	4.1 GenomeAxisTrack	1.1			
	4.2 IdeogramTrack	. 14			
	4.3 DataTrack	15			
	4.4 AnnotationTrack	. 28			
	4.5 GeneRegionTrack	3.			
	4.6 BiomartGeneRegionTrack	39			
	4.7 DetailsAnnotationTrack	41			
	4.8 SequenceTrack	46			
	4.9 Creating tracks from UCSC data	48			
5	Bioconductor integration and file support	<b>5</b> 1			
6	Composite plots for multiple chromosomes				

# 1 Introduction

In order to make sense of genomic data one often aims to plot such data in a genome browser, along with a variety of genomic annotation features, such as gene or transcript models, CpG island, repeat regions, and so on. These features may either be extracted from public data bases like ENSEMBL or UCSC, or they may be generated or curated in-house. Many of the currently available genome browsers do a reasonable job in displaying genome annotation data, and there are options to connect to some of them from within R (e.g., using the rtracklayer package). However, none of these solutions offer the flexibility of the full R graphics system to display large numeric data in a multitude of different ways. The Gviz package aims to close this gap by providing a structured visualization framework to plot any type of data along genomic coordinates. It is loosely based on the GenomeGraphs package by Steffen Durinck and James Bullard, however the complete class hierarchy as well as all the plotting methods have been restructured in order to increase performance

<sup>\*</sup>florian.hahne@novartis.com

and flexibility. All plotting is done using the grid graphics system, and several specialized annotation classes allow to integrate publicly available genomic annotation data from sources like UCSC or ENSEMBL.

# 2 Basic Features

The fundamental concept behind the Gviz package is similar to the approach taken by most genome browsers, in that individual types of genomic features or data are represented by separate tracks. Within the package, each track constitutes a single object inheriting from class GdObject, and there are constructor functions as well as a broad range of methods to interact with and to plot these tracks. When combining multiple objects, the individual tracks will always share the same genomic coordinate system, thus taking the burden of aligning the plot elements from the user.

It is worth mentioning that, at a given time, tracks in the sense of the Gviz package are only defined for a single chromosome on a specific genome, at least during the plotting operation. You will later see that a track may still contain information for multiple chromosomes, however most of this is hidden except for the currently active chromosome, and the user will have to explicitly switch the chromosome to access the inactive parts. While the package in principle imposes no fixed structure on the chromosome or on the genome names, it makes sense to stick to a standaradized naming paradigm, in particular when fetching additional annotation information from online resources. By default this is enforced by a global option ucscChromosomeNames, which is set during package loading and which causes the package to check all supplied chromosome names for validity in the sense of the UCSC definition (chromosomes have to start with the *chr* string). You may decide to turn this feature off by calling options(ucscChromosomeNames=FALSE). For the remainder of this vignette however, we will make use of the UCSC genome and chromosome identifiers, e.g., the chr7 chromosome on the mouse mm9 genome.

The different track classes will be described in more detail in the Track classes section further below. For now, let's just take a look at a typical Gviz session to get an idea of what this is all about. We begin our presentation of the available functionality by loading the package:

#### > library(Gviz)

The most simple genomic features consists of start and stop coordinates, possibly overlapping each other. CpG islands or microarray probes are real life examples for this class of features. In the Bioconductor world those are most often represented as run-length encoded vectors, for instance in the *IRanges* and *GRanges* classes. To seamlessly integrate with other Bioconductor packages, we can use the same data structures to generate our track objects. A sample set of CpG island coordinates has been saved in the cpgIslands object and we can use that for our first annotation track object. The constructor function AnnotationTrack is a convenient helper to create the object.

```
> library(GenomicRanges)
> data(cpgIslands)
> class(cpgIslands)

[1] "GRanges"
attr(,"package")
[1] "GenomicRanges"

> chr <- as.character(unique(seqnames(cpgIslands)))
> gen <- genome(cpgIslands)
> atrack <- AnnotationTrack(cpgIslands, name = "CpG")</pre>
```

Please note that the AnnotationTrack constructor (as most constructors in this package) is fairly flexible and can accommodate many different types of inputs. For instance, the start and end coordinates of the annotation features could be passed in as individual arguments start and end, as a data.frame or even as an IRanges or GRangesList object. Furthermore, a whole bunch of coercion methods are available for the more traditional package users that should allow operations along the lines of as(obj, 'AnnotationTrack'). You may want to consult the class' manual page for more information, or take a look at the Bioconductor integration section for a listing of the most common data structures and their respective counterparts in the Gviz package.

With our first track object being created we may now proceed to the plotting. There is a single function plotTracks that handles all of this. As we will learn in the remainder of this vignette, plotTracks is quite powerful and has a number of very useful additional arguments. For now we will keep things very simple and just plot the single CpG islands annotation track.

#### > plotTracks(atrack)



As you can see, the resulting graph is not particularly spectacular. There is a title region showing the track's name on a gray background on the left side of the plot and a data region showing the seven individual CpG islands on the right. This structure is similar for all the available track objects classes and it somewhat mimicks the layout of the popular UCSC Genome Browser. If you are not happy with the default settings, the Gviz package offers a multitude of options to fine-tune the track appearance, which will be shown in the Plotting Parameters section.

Appart from the relative distance of the Cpg islands, this visualization does not tell us much. One obvious next step would be to indicate the genomic coordinates we are currently looking at to provide some reference. For this purpose, the Gviz package offers the *GenomeAxisTrack* class. Objects from the class can be created using the constructor function of the same name.

# > gtrack <- GenomeAxisTrack()</pre>

Since a *GenomeAxisTrack* object is always relative to the other tracks that are plotted, there is little need for additional arguments. Essentially, the object just tells the plotTracks function to add a genomic axis to the plot. Nonetheless, it represent a separate annotation track just as the CpG island track does. We can pass this additional track on to plotTracks in the form of a list.

# > plotTracks(list(gtrack, atrack))



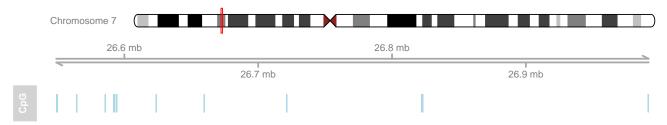
You may have realized that the genomic axis does not take up half of the available vertical plotting space, but only uses the space necessary to fit the axis and labels. Also the title region for this track is empty. In general, the Gviz package tries to find reasonable defaults for all the parameters controlling the look and feel of a plots so that apealing visualizations can be created without much tinkering. However, all features on the plot including the relative track sizes can also be adjusted manually.

As mentioned before in the beginning of this vignette, a plotted track is always defined for exactly one chromosome on a particular genome. We can include this information in our plot by means of a chromosome ideogram. An ideogram is a simplified visual representation of a chromosome, with the different chromosomal staining bands indicated by color, and the centromer (if present) indicated by the shape. The necessary information to produce this visualization is stored in online data repositories, for instance at UCSC. The Gviz package offers very convenient connections to some of these repositories, and the *IdeogramTrack* constructor function is one example for such a connection. With just the information about a valid UCSC genome and chromosome, we can directly fetch the chromosome ideogram information and construct a dedicated track object that can be visualized by plotTracks. Please not that you will need an established internet connection for this to work, and that fetching data from UCSC can take quite a long time, depending on the server load. The Gviz package tries to cache as much data as possible to reduce the bandwidth in future queries.

# > itrack <- IdeogramTrack(genome = gen, chromosome = chr)</pre>

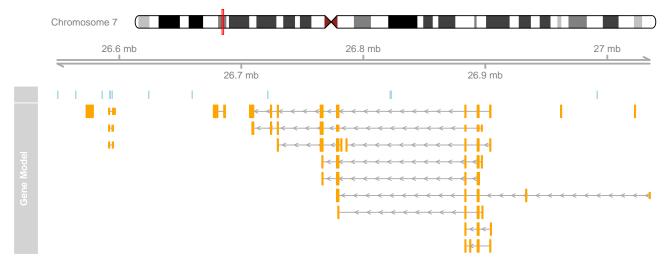
Similar to the previous examples, we stick the additional track object into a list in order to plot it.

# > plotTracks(list(itrack, gtrack, atrack))



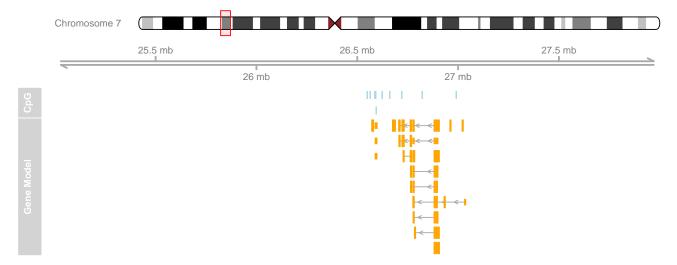
Ideogram tracks are the one exception in all of Gviz 's track objects in the sense that they are not really displayed on the same coordinate system like all the other tracks. Instead, the current genomic location is indicated on the chromosome by a red box (or, as in this case, a red line if the width is too small to fit a box). So far we have only looked at very basic annotation features and how to give a point of reference to our plots. Naturally, we also want to be able to handle more complex genomic features, such as gene models. One potential use case would be to utilize gene model information from an existing local source. Alternatively, we could dowload such data from one of the available online resources like UCSC or ENSEBML, and there are constructor functions to handle these tasks. For this example we are going to load gene model data from a stored data frame. The track class of choice here is a GeneRegionTrack object, which can be created via the constructor function of the same name. Similar to the AnnotationTrack constructor there are multiple possible ways to pass in the data.

- > data(geneModels)
- > grtrack <- GeneRegionTrack(geneModels, genome = gen,
- + chromosome = chr, name = "Gene Model")
- > plotTracks(list(itrack, gtrack, atrack, grtrack))



So far the plotted genomic range has been determined by the input tracks. Unless told otherwise, the package will always display the region from the leftmost item to the rightmost item in any of the tracks. Of course such a static view on a chromosomal region is of rather limited use. We often want to zoom in or out on a particular plotting region to see more details or to get a broader overview. To that end, plotTracks supports the from and to arguments that let us choose an arbitrary genomic range to plot.

- > plotTracks(list(itrack, gtrack, atrack, grtrack),
- + from = 25000000, to = 28000000)

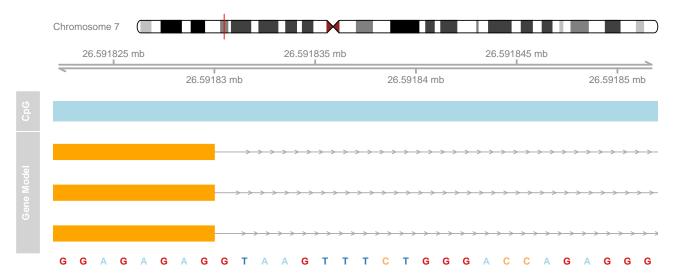


You may have noticed that the layout of the gene model track has changed depending on the zoom level. This is a feature of the Gviz package, which automatically tries to find the optimal visualization settings to make best use of the available space. At the same time, when features on a track are too close together to be plotted as separate items with the current device resolution, the package will try to reasonably merge them in order to avoid overplotting.

When zooming further in it may become intersting to take a look at the actual genomic sequence at a given position, and the Gviz package provides the track class SequenceTrack that let's you do that. It can draw the necessary sequence information from one of the BSqenome packages.

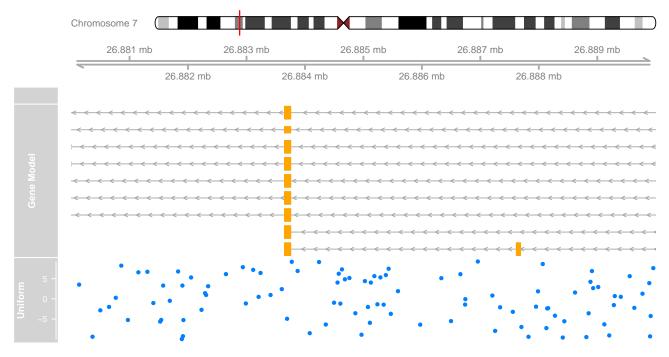
- > library(BSgenome.Hsapiens.UCSC.hg19)
- > strack <- SequenceTrack(Hsapiens, chromosome = chr)

```
> plotTracks(list(itrack, gtrack, atrack, grtrack,
+ strack), from = 26591822, to = 26591852, cex = 0.8)
```



So far we have replicated the features of a whole bunch of other genome browser tools out there. The real power of the package comes with a rather general track type, the DataTrack. DataTrack object are essentially run-length encoded numeric vectors or matrices, and we can use them to add all sorts of numeric data to our genomic coordinate plots. There are a whole bunch of different visualization options for these tracks, from dot plots to histograms to box-and-whisker plots. The individual rows in a numeric matrix are considered to be different data groups or samples, and the columns are the raster intervals in the genomic coordinates. Of course, the data points (or rather the data ranges) do not have to be evenly spaced; each column is associated with a particular genomic location. For demonstration purposes we can create a simple DataTrack object from randomly sampled data.

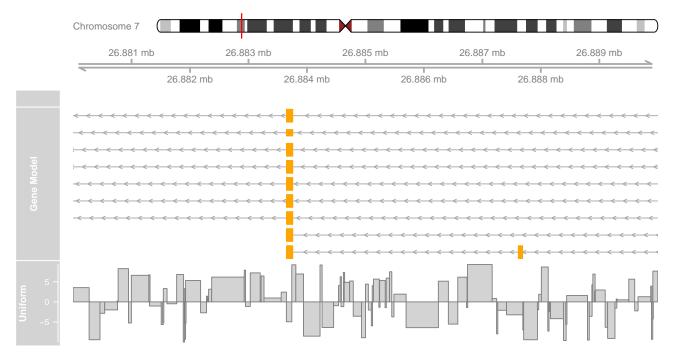
```
> set.seed(255)
> lim <- c(26880000, 26890000)
> coords <- sort(c(lim[1], sample(seq(from = lim[1],
+ to = lim[2]), 99), lim[2]))
> dat <- runif(100, min = -10, max = 10)
> dtrack <- DataTrack(data = dat, start = coords[-length(coords)],
+ end = coords[-1], chromosome = chr, genome = gen,
+ name = "Uniform")
> plotTracks(list(itrack, gtrack, atrack, grtrack,
+ dtrack), from = lim[1], to = lim[2])
```



The first thing to notice is that the title panel to the right of the plot now contains a y-axis indicating the range of the displayed data. The default plotting type for numeric vectors is a simple dot plot. This is by far not the only visualization option, and in a sense it is waisting quite a lot of information because the run-length encoded ranges are not immediately apparent. We can change the plot type by supplying the type argument to plotTracks. A complete description of the available plotting options is given in section Track classes, and a more detailed treatment of the so-called 'display parameters' that control the look and feel of a track is given in the Plotting Parameters section.

```
> plotTracks(list(itrack, gtrack, atrack, grtrack,
```

<sup>+</sup> dtrack), from = lim[1], to = lim[2], type = "histogram")



As we can see, the data values in the numeric vector are indeed matched to the genomic coordinates of the DataTrack object. Such a visualization can be particularly helpful when displaying for instance the coverage of NGS reads along a chromosome, or to show the measurement values of mapped probes from a micro array experiment.

This concludes our first introduction into the Gviz package. The remainder of this vignette will deal in much more depth with the different concepts and the various track classes and plotting options.

# 3 Plotting parameters

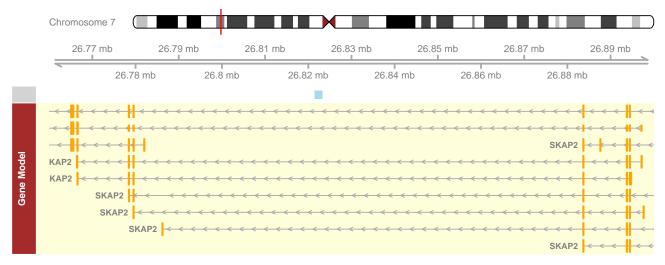
\$min.distance

Although not implicitely mentioned before, we have already made use of the plotting parameter facilities in the Gviz package, or, as we will call them from now on, the 'display parameters'. Display parameters are properties of individual track objects (i.e., of any object inheriting from the base GdObject class). They can either be set during object instantiation as additional arguments to the constructor functions or, for existing track objects, using the displayPars replacement method. In the former case, all named arguments that can not be matched to any of the constructor's formal arguments are considered to be display parameters, regardless of their type or whether they are defined for a particular track class or not. The following code example rebuilds our GeneRegionTrack object with a bunch of display parameters and demonstrates the use of the displayPars accessor and replacement methods.

```
> grtrack <- GeneRegionTrack(geneModels, genome = gen,
+ chromosome = chr, name = "Gene Model", showId = TRUE,
+ background.title = "brown")
> head(displayPars(grtrack))

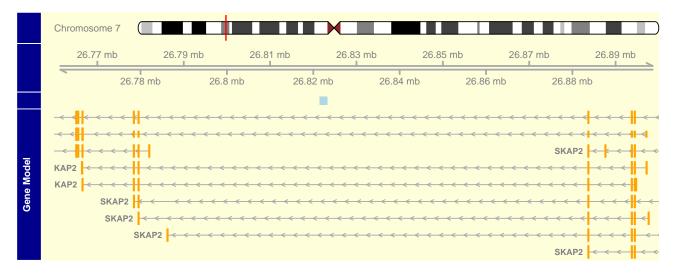
$fill
[1] "orange"
```

```
[1] 0
$col
NULL
$geneSymbols
[1] TRUE
$showExonId
[1] FALSE
$collapseTranscripts
[1] FALSE
> displayPars(grtrack) <- list(background.panel = "#FFFEDB")</pre>
> head(displayPars(grtrack))
$fill
[1] "orange"
$min.distance
[1] 0
$col
NULL
$geneSymbols
[1] TRUE
$showExonId
[1] FALSE
$collapseTranscripts
[1] FALSE
> plotTracks(list(itrack, gtrack, atrack, grtrack),
      from = lim[1] - 120000, to = lim[2] + 10000)
```



For our gene model track we have now added the gene symbols of the different transcripts to the plot and changed the background color of both the title and the data panel. There is a third option to set display parameters for a single plotting operation (rather than the permanent setting in the track object) by passing in additional arguments to the plotTracks function. We have already made use of this feature in the previous data plotting type example. It is worth mentioning that all display parameters which are passed along with the plotTracks function apply to all track objects in the plot. For some objects classes a particular display parameter may not make any sense, and in that case it is simply ignored. Also, the settings only apply for one single plotting operation and will not be retained in the plotted track objects. They do however get precedence over the object-internal parameters. The following line of code exemplifies this behaviour.

- > plotTracks(list(itrack, gtrack, atrack, grtrack),
- + from = lim[1] 120000, to = lim[2] + 10000,
- + background.panel = "#FFFEDB", background.title = "darkblue")



In order to make full use of the flexible parameter system we need to know which display parameters control which aspect of which track class. The obvious source for this information are the man pages of the respective track classes, which list all available parameters along with a short description of their effect and default values in the Display Parameters section. Alternatively, we can use the availableDisplayPars function, which prints out the available parameters for a class as well as their default values in a list-like

structure. The single argument to the function is either a class name of a track object class, or the object itself, in which case its class is automatically detected.

```
> dp <- availableDisplayPars(grtrack)
> tail(dp)
The following display parameters are available for 'GeneRegionTrack' objects:
(see ? GeneRegionTrack for details on their usage)
shape: smallArrow box
showAxis (inherited from class 'GdObject'): TRUE
showExonId: FALSE
showFeatureId (inherited from class 'AnnotationTrack'): FALSE
showId (inherited from class 'AnnotationTrack'): FALSE
showOverplotting (inherited from class 'AnnotationTrack'): FALSE
showTitle (inherited from class 'GdObject'): TRUE
size (inherited from class 'GdObject'): 1
stackHeight (inherited from class 'StackedTrack'): 0.75
thinBoxFeature: utr ncRNA utr3 utr5 3UTR 5UTR miRNA lincRNA three_prime_UTR five_prime_UTR
v (inherited from class 'GdObject'): -1
```

As we can see, display parameters can be inherited from parent classes. For the regular user this is not important at all, however it nicely exemplifies the structure of the class hierarchy in the Gviz package.

# 4 Track classes

In this section we will highlight all of the available annotation track classes in the Gviz package. For the complete reference of all the nuts and bolts, including all the available methods, please see the respective class man pages. We will try to keep this vignette up to date, but in cases of discrepancies between here and the man pages you should assume the latter to be correct.

#### 4.1 GenomeAxisTrack

GenomeAxisTrack objects can be used to add some reference to the currently displayed genomic location to a Gviz plot. In their most basic form they are really just a horizontal axis with genomic coordinate tickmarks. Using the GenomeAxisTrack constructor function is the recommended way to instantiate objects from the class. There is no need to know in advance about a particular genomic location when constructing the object. Instead, the displayed coordinates will be determined from the context, e.g., from the from and to arguments of the plotTracks function, or, when plotted together with other track objects, from their genomic locations.

As an optional feature one can highlight particular regions on the axis, for instance to indicated stretches of N nucleotides or gaps in genomic alignments. Such regions have to be supplied to the optional *range* argument of the constructor function as either an *GRanges* or an *IRanges* object.

- > axisTrack <- GenomeAxisTrack(range = IRanges(start = c(2000000,
- + 4000000), end = c(3000000, 7000000), names = rep("N-stretch",
- + 2)))
- > plotTracks(axisTrack, from = 1000000, to = 9000000)



If names have been supplied with the range argument, those can also be added to the plot.

- > plotTracks(axisTrack, from = 1000000, to = 9000000,
- + showId = TRUE)



### Display parameters for GenomeAxisTrack objects

There are a whole bunch of display parameters to alter the appearance of *GenomeAxisTrack* plots. For instance, one could add directional indicators to the axis using the *add53* and *add35* parameters.

- > plotTracks(axisTrack, from = 1000000, to = 9000000,
- + add53 = TRUE, add35 = TRUE)



Sometimes the resolution of the tick marks is not sufficient, in which case the *littleTicks* argument can be used to have a more fine-grained axis annotation.

- > plotTracks(axisTrack, from = 1000000, to = 9000000,
- + add53 = TRUE, add35 = TRUE, littleTicks = TRUE)



The Gviz package tries to come up with reasonable defaults for the axis annotation. In our previous example, the genomic coordinates are indicated in megabases. We can control this via the *exponent* parameter, which takes an integer value greater then zero. The location of the tick marks are displayed as a fraction of  $10^{exponent}$ .

- > plotTracks(axisTrack, from = 1000000, to = 9000000,
- + exponent = 4)



Another useful parameter, *labelPos* controls the arrangement of the tick marks. It takes one of the values alternating, revAlternating, above or below. For instance we could aline all tick marks underneath the axis.

```
> plotTracks(axisTrack, from = 1000000, to = 9000000,
+ labelPos = "below")
```



Sometimes a full-blown axis is just too much, and all we really need in the plot is a small scale indicator. We can change the appearance of the *GenomeAxisTrack* object to such a limited representation by setting the *scale* display parameter. Typically, this will be a numeric value between 0 and 1, which is interpreted as the fraction of the plotting region used for the scale. The plotting method will apply some rounding to come up with reasonable and human-readable values. For even more control we can pass in a value larger than 1, which is considered to be an absolute range length. In this case, the user is responsible for the scale to actually fit in the current plotting range.

We still have control over the placement of the scale label via the *labelPos*, parameter, which now takes the values above, below and beside.

```
> plotTracks(axisTrack, from = 1000000, to = 9000000, 
+ scale = 0.5, labelPos = "below")
```

For a complete listing of all the available display parameters please see the table below or the man page of the GenomeAxisTrack class by typing in ?GenomeAxisTrack on the R command line.

Display Parameter	Description
add35	Logical scalar. Add 3' to 5' direction indicators.
add53	Logical scalar. Add 5' to 3' direction indicators.
cex	Numeric scalar. The overall font expansion factor for the axis annotation
	text.
cex.id	Numeric scalar. The text size for the optional range annotation.
col	Character scalar. The color for the axis lines and tickmarks.
col.id	Character scalar. The text color for the optional range annotation.
col.range	Character scalar. The border color for highlighted regions on the axis.
distFromAxis	Numeric scalar. Control the distance of the axis annotation from the tick marks.
exponent	Numeric scalar. The exponent for the axis coordinates, e.g., 3 means mb, 6 means gb, etc. The default is to automatically determine the optimal exponent.
fill.range	Character scalar. The fill color for highlighted regions on the axis.
fontcolor	Character scalar. The font color for the axis annotation text.
fontface	Character scalar. The font face for the axis annotation text.
fontfamily	Character scalar. The font family for the axis annotation text.

fontsize Numeric scalar. Font size for the axis annotation text in points.

Character vector, one in "alternating", "revAlternating", "above" or "below".

The vertical positioning of the axis labels. If scale is not NULL, the possible values are "above", "below" and "beside".

Logical scalar. Add more fine-grained tick marks.

Numeric scalar. The line width for the axis elementes.

Scale Numeric scalar. If not NULL a small scale is drawn instead of the full axis, if the value is between 0 and 1 it is interpreted as a fraction of the current plotting region, otherwise as an absolute length value in genomic coordinates. showId Logical scalar. Show the optional range highlighting annotation.

#### 4.2 IdeogramTrack

While a genomic axis provides helpful points of reference to a plot, it is sometimes important to show the currently displayed region in the broader context of the whole chromosme. Are we looking at distal regions, or somewhere close to the centromer? And how much of the complete chromosome is covered in our plot. To that end the Gviz package defines the *IdeogramTrack* class, which is an idealized representation of a single chromosome. When plotted, these track objects will always show the whole chromosome, regardless of the selected genomic region. However, the displayed coordinates are indicated by a box that sits on the ideogram image. The chromosomal data necessary to draw the ideogram is not part of the Gviz package itself, instead it is downloaded from an online source (UCSC). Thus it is important to use both chromosome and genome names that are recognizable in the UCSC data base when dealing with *IdeogramTrack* objects. You might want to consult the UCSC webpage (http://genome.ucsc.edu/) or use the ucscGenomes function in the rtracklayer package for a listing of available genomes.

Assuming the chromosome data are available online, a simple call to the IdeogramTrack constructor function including the desired genome and chromosome name are enough to instantiate the object. Since the connection to UCSC can be slow, the package tries to cache data that has already been downloaded for the duration of the R session. If needed, the user can manually clear the cache by calling the clearSessionCache function. Of course it is also possible to construct *IdeogramTrack* objects from local data. Please see the class' man page for details.

```
> ideoTrack <- IdeogramTrack(genome = "hg19", chromosome = "chrX")
> plotTracks(ideoTrack, from = 85000000, to = 129000000)
Chromosome X
```

We can turn off the explicit plotting of the chromosome name by setting the showId display parameter to FALSE.

```
> plotTracks(ideoTrack, from = 85000000, to = 129000000,
+ showId = FALSE)
```

The chromosome bands in the ideogram come with a unique identifier, and we can add this information to the plot, at least for those bands that are wide enought to accommodate the text.

```
> plotTracks(ideoTrack, from = 85000000, to = 129000000,
+ showId = FALSE, showBandId = TRUE, cex.bands = 0.5)
```



#### Display parameters for IdeogramTrack objects

For a complete listing of all the available display parameters please see the table below or the man page of the *IdeogramTrack* class by typing in ?IdeogramTrack on the R command line.

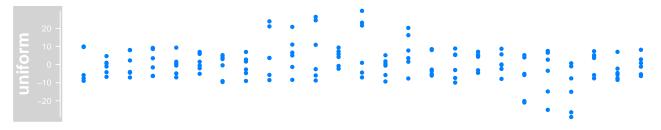
Display Parameter	Description
bevel	Numeric scalar, between 0 and 1. The level of smoothness for the two ends
	of the ideogram.
cex	Numeric scalar. The overall font expansion factor for the chromosome name
	text.
cex.bands	Numeric scalar. The font expansion factor for the chromosome band identi-
	fier text.
col	Character scalar. The border color used for the highlighting of the currently
	displayed genomic region.
fill	Character scalar. The fill color used for the highlighting of the currently
	displayed genomic region.
fontcolor	Character scalar. The font color for the chromosome name text.
fontface	Character scalar. The font face for the chromosome name text.
fontfamily	Character scalar. The font family for the chromosome name text.
fontsize	Numeric scalar. The font size for the chromosome name text.
lty	Character or integer scalar. The line type used for the highlighting of the
	currently displayed genomic region.
lwd	Numeric scalar. The line width used for the highlighting of the currently
	displayed genomic region.
showBandId	Logical scalar. Show the identifier for the chromosome bands if there is space
	for it.
showId	Logical scalar. Indicate the chromosome name next to the ideogram.

#### 4.3 DataTrack

Probably the most powerfull of all the track classes in the Gviz package are DataTracks. Essentially they constitute run-length encoded numeric vectors or matrices, meaning that one or several numeric values are associated to a particular genomic coordinate range. These ranges may even be overlapping, for instance when looking at results from a running window operation. There can be multiple samples in a single data set, in which case the ranges are associated to the columns of a numeric matrix rather than a numeric vector, and the plotting method provides tools to incoorporate sample group information. Thus the starting point for creating DataTrack objects will always be a set of ranges, either in the form of an IRanges or GRanges object, or individually as start and end coordinates or widths. The second ingredient is a numeric vector of the same length as the number of ranges, or a numeric matrix with the same number of columns. Those may even already be part of the input GRanges object as elemenMetadata values. For a complete description of all the possible inputs please see the class' online documentation. We can pass all this information to the DataTrack constructor function to instantiate an object. We will load our sample data from an GRanges

object that comes as part of the Gviz package.

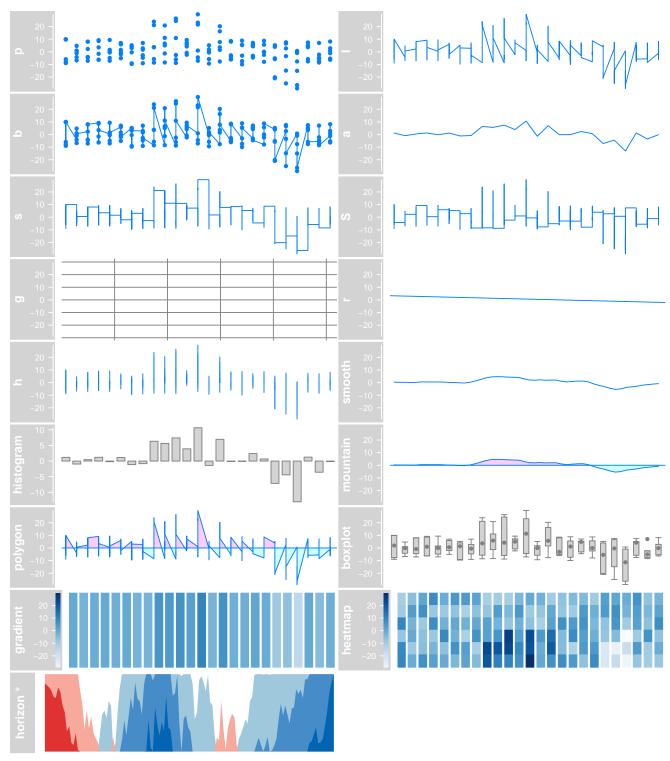
- > data(twoGroups)
- > dTrack <- DataTrack(twoGroups, name = "uniform")</pre>
- > plotTracks(dTrack)



The default visualization for our very simplistic sample *DataTrack* is a rather unispiring dot plot. The track comes with a scale to indicate the range of the numeric values on the y-axis, appart from that it looks very much like the previous examples. A whole battery of display parameters is to our disposal to control the track's look and feel. The most important one is the *type* parameter. It determines the type of plot to use and takes one or several of the following values:

Value	Type
p	dot plot
1	lines plot
b	dot and lines plot
a	lines plot of average (i.e., mean) values
$\mathbf{s}$	stair steps (horizontal first)
$\mathbf{S}$	stair steps (vertical first)
g	add grid lines
$\mathbf{r}$	add linear regression line
h	histogram lines
$\operatorname{smooth}$	add loess curve
histogram	histogram (bar width equal to range with)
mountain	'mountain-type' plot relative to a baseline
polygon	'polygon-type' plot relative to a baseline
boxplot	box and whisker plot
gradient	false color image of the summarized values
heatmap	false color image of the individual values
horizon	Horizon plot indicating magnitude and direction of a change relative to a
	baseline

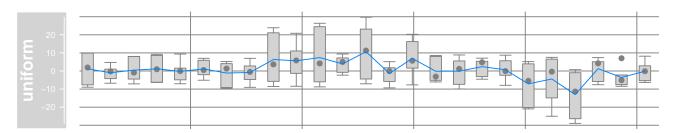
Displayed below are the same sample data as before but plotted with the different type settings:



You will notice that some of the plot types work better for univariate data while others are clearly designed for multivariate inputs. The *a* type for instance averages the values at each genomic location before plotting the derived values as a line. The decision for a particular plot type is totally up to the user, and one could even overlay multiple types by supplying a character vector rather than a character scalar as the (type) argument. For example, this will combine a boxplot with an average line and a data grid.

<sup>\*</sup>A different data set is plotted for the horizon type for the sake of clarity.

> plotTracks(dTrack, type = c("boxplot", "a", "g"))



For the heatmap plotting type we arrange all the data in a well-structured two-dimensional matrix which gives us the oportunity to add a little extra information about the individual samples. Depending on how the *DataTrack* was created in the first place we can choose to display the sample names (which in our case correspond to the column names of the input *GRanges* object). The plot also highlights another feature of the heatmap type: the y-axis now shows a mapping of the numeric values into the color range.

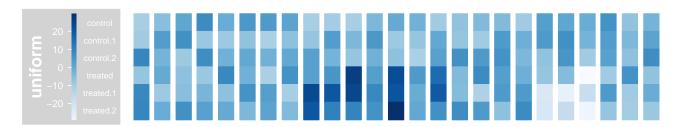
> colnames(mcols(twoGroups))

```
[1] "control" "control.1" "control.2" "treated.1"
```

[6] "treated.2"

> plotTracks(dTrack, type = c("heatmap"), showSampleNames = TRUE,

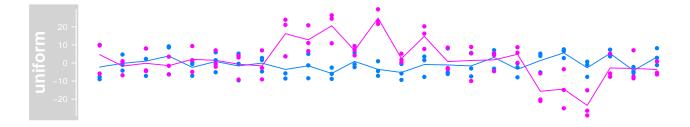
+ cex.sampleNames = 0.6)



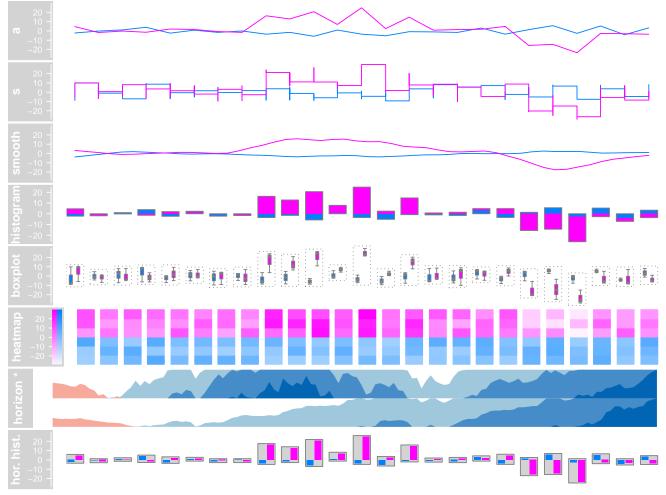
# **Data Grouping**

An additional layer of flexibility is added by making use of Gviz 's grouping functionality. The individual samples (i.e., rows in the data matrix) can be grouped together using a factor variable, and, if reasonable, this grouping is reflected in the layout of the respective track types. For instance our example data could be derived from two different sample groups with three replicates each, and we could easily integrate this information into our plot.

```
> plotTracks(dTrack, groups = rep(c("control", "treated"),
+ each = 3), type = c("a", "p"))
```



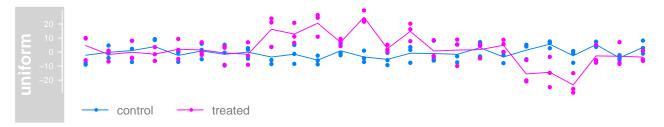
For the dot plot representation the individual group levels are indicated by color coding. For the a type, the averages are now computed for each group separately and also indicated by two lines with similar color coding. Grouping is not supported for all plotting types, for example the mountain and polygon type already use color coding to convey a different message and for the gradient type the data are already collapsed to a single variable. The following gives an overview over some of the other groupable DataTrack types. Please note that there are many more display parameters that control the layout of both grouped and of ungrouped DataTracks. You may want to check the class' help page for details.



If we need to display some additional information about the individual group levels we can make use of the *legend* display parameter to add a simple legend to the plot. Depending on the plot type and on some of the other display parameters, the look of this legend may vary slightly.

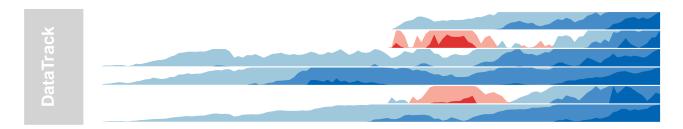
```
> plotTracks(dTrack, groups = rep(c("control", "treated"),
+ each = 3), type = c("a", "p"), legend = TRUE)
```

<sup>\*</sup>A different data set is plotted for the horizon type for the sake of clarity.



For a grouped horizon plot the group labels have to be shown in a similar fashion as for heatmaps, i.e., by setting the *showSampleNames* argument to TRUE.

```
> data(dtHoriz)
> dtHoriz <- dtHoriz[1:6, ]
> plotTracks(dtHoriz, type = "horiz", groups = rownames(values(dtHoriz)),
+ showSampleNames = TRUE, cex.sampleNames = 0.6,
+ title.width = 2, separator = 1)
```



#### Building DataTrack objects from files

A number of standard file types exist that all store numeric data along genomic coordinates. We have tried to make such files accessible in the Gviz package by providing additional options to the DataTrack constructor function. In the previous examples the range argument was a GRanges object. Instead, we can also pass in the path to a file on disk by means of a character scalar. The DataTrack class supports the most common file types like wig, bigWig or bedGraph, but also knows how to deal with bam files. You may have realized that some of these files are indexed, and we have taken the approach to stream the data from indexed files on the fly when it is needed for plotting.

However let's first start with the simple example of a bedGraph file. These files support a single data sample, and thus are equivalent to a GRanges object with a single numeric elementMetadata column. bedGraph files are not indexed, so we have to load the whole file content when instantiating the object.

```
> bgFile <- system.file("extdata/test.bedGraph", package = "Gviz")
> dTrack2 <- DataTrack(range = bgFile, genome = "hg19",
+ type = "l", chromosome = "chr19", name = "bedGraph")
> class(dTrack2)

[1] "DataTrack"
attr(,"package")
[1] "Gviz"
> plotTracks(dTrack2)
```

# pedGraph - 0.5 - - 0.0 - - 0.5 - - 0.1

As we can see the constructor has returned a regular DataTrack object. The function to be used in order to read the data off the file has been automatically choosen by the package based on the file extension of the input file. Of course the number of these supported standard file types is limited, and a user may want to import a non-standard file through the same mechanism. To this end, the DataTrack constructor defines an additional argument called importFunction. As the name suggests, the value of this argument is a function which needs to handle the mandatory file argument. Upon evaluation this argument will be filled in with the path to the data file, and the user-defined function needs to provide all logic necessary to parse that file into a valid GRanges object. From this point on everything will happen just as if the range argument had been this GRanges object. In other words, numeric elementMetadata columns will be shown as individual samples and non-numeric columns will be silently ignored. We can exemplify this in the next code chunk. Note that the Gviz package is using functionality from the rtracklayer package for most of the file import operations, just as we do here in a more explicit way.

```
> library(rtracklayer)
> dTrack3 <- DataTrack(range = bgFile, genome = "hg19",
+ type = "l", chromosome = "chr19", name = "bedGraph",
+ importFunction = function(file) import(con = file,
+ asRangedData = FALSE))
> identical(dTrack2, dTrack3)
```

#### [1] TRUE

So far one could have easily done the whole process in two separate steps: first import the data from the file into a *GRanges* object and then provided this object to the constructor. The real power of the file support in the Gviz package comes with streaming from indexed files. As mentioned before, only the relevant part of the data has to be loaded during the plotting operation, so the underlying data files may be quite large without decreasing the performance or causing too big of a memory footprint. We will exemplify this feature here using a small bam file that is provided with the package. bam files contain alignments of sequences (typically from a next generation sequencing experiment) to a common reference. The most natural representation of such data in a *DataTrack* is to look at the alignment coverage at a given position only and to encode this in a single elementMetadata column.

```
> bamFile <- system.file("extdata/test.bam", package = "Gviz")
> dTrack4 <- DataTrack(range = bamFile, genome = "hg19",
+ type = "l", name = "Coverage", window = -1, chromosome = "chr1")
> class(dTrack4)

[1] "ReferenceDataTrack"
attr(,"package")
[1] "Gviz"
> dTrack4
```

```
ReferenceDataTrack 'Coverage'
| genome: hg19
| active chromosome: chr1
| referenced file: /tmp/RtmpguvUpW/Rinst2c267c108d3e/Gviz/extdata/test.bam
> plotTracks(dTrack4, from = 189990000, to = 190000000)
```



As seen in the previous code chunk, the dTrack4 object is now of class ReferenceDataTrack. For the user this distinction is not particularly relevant with the exception that the length method for this class almost always returns 0 because the content of the object is only realized during the plotting operation. Obviously, streaming from the disk comes with a price in that file access is much slower than accessing RAM, however the file indexing allows for fairly rapid data retrieval, and other processes during the plotting operation tend to be much more costly, anyways. It is woth mentioning however that each plotting operation will cause reading off the file, and there are currently no caching mechanisms in place to avoid that. Nevertheless, plotting a larger chunk of the bam file still finishes in a reasonable time.

```
> plotTracks(dTrack4, chromosome = "chr1", from = 189891483,
+ to = 190087517)
```

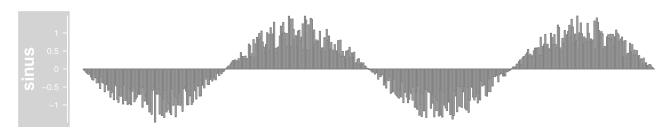
Of course users can provided their own file parsing function just like we showed in the previous example. The import function now needs to be able to deal with a second mandatory argument *selection*, which is a *GRanges* object giving the genomic interval that has to be imported from the file. In addition one needs to tell the *DataTrack* constructor that data should be streamed off a file by setting the *stream* argument to TRUE.

#### **Data transformations**

The Gviz package offers quite some flexibility to transform data on the fly. This involves both rescaling operations (each data point is transformed on the track's y-axis by a transformation function) as well as

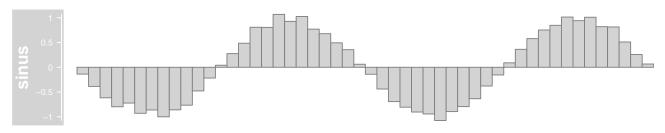
summarization and smoothing operations (the values for several genomic locations are summarized into one derived value on the track's x-axis). To illustrate this let's create a significantly bigger DataTrack than the one we used before, containing purely syntetic data for only a single sample.

```
> dat <- sin(seq(pi, 10 * pi, len = 500))
> dTrack.big <- DataTrack(start = seq(1, 100000, len = 500),
+ width = 15, chromosome = "chrX", genome = "hg19",
+ name = "sinus", data = sin(seq(pi, 5 * pi, len = 500)) *
+ runif(500, 0.5, 1.5))
> plotTracks(dTrack.big, type = "hist")
```



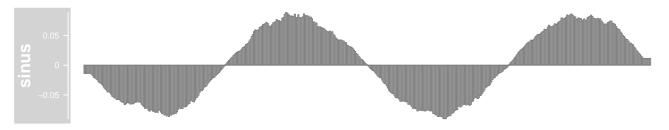
Since the available resolution on our screen is limited we can no longer distinguish between individual coordinate ranges. The Gviz package tries to avoid overplotting by collapsing overlapping ranges (assuming the collapse Tracks parameter is set to TRUE). However, it is often desirable to summarize the data, for instance by binning values into a fixed number of windows followe by the calculation of a meaningful summary statistic. This can be archived by a combination of the window and aggregation display parameters. The former can be an integer value greater than zero giving the number of evenly-sized bins to aggregate the data in. The latter is supposed to be a user-supplied function that accepts a numeric vector as a single input parameter and returns a single aggregated numerical value. For simplicity, the most obvious aggregation functions can be selected by passing in a character scalar rather than a function. Possible values are mean, median, extreme, sum, min and max. These presets are also much faster because they have been optimized to operate on large numeric matrices. The default is to compute the mean value of all the binned data points.

#### > plotTracks(dTrack.big, type = "hist", window = 50)



Instead of binning the data in fixed width bins one can also use the *window* parameter to perform more elaborate running window operations. For this to happen the parameter value has to be smaller than zero, and the additional display parameter *windowSize* can be used to control the size of the running window. This operation does not change the number of coordinate ranges on the plot, but instead the original value at a particular position is replaced by the respective sliding window value at the same position. A common use case for sliding windows on genomic ranges is to introduce a certain degree of smoothing to the data.

```
> plotTracks(dTrack.big, type = "hist", window = -1,
+ windowSize = 2500)
```



In addition to transforming the data on the x-axis we can also apply arbitrary transformation functions on the y-axis. One obvious use-case would be to log-transform the data prior to plotting. The framework is flexible enough however to allow for arbitrary transformation operations. The mechanism works by providing a function as the *transformation* display parameter, which takes as input a numeric vector and returns a transformed numeric vector of the same length. The following code for instance truncates the plotted data to values greater than zero.

```
> plotTracks(dTrack.big, type = "l", transformation = function(x) {
+    x[x < 0] <- 0
+    x
+ })</pre>
```



As seen before, the a type allows to plot average values for each of the separate groups. There is however an additional parameter aggregate Groups that generalizes group value aggregations. In the following example we display, for each group and at each position, the average values in the form of a dot-and-lines plot.

```
> plotTracks(dTrack, groups = rep(c("control", "treated"),
+ each = 3), type = c("b"), aggregateGroups = TRUE)
```



This functionality again also relies on the setting of the *aggregation* parameter, and we can easily change it to display the maximum group values instead.

```
> plotTracks(dTrack, groups = rep(c("control", "treated"),
+ each = 3), type = c("b"), aggregateGroups = TRUE,
+ aggregation = "max")
```



# Display parameters for DataTrack objects

For a complete listing of all the available display parameters please see the table below or the man page of the DataTrack class by typing in <code>?DataTrack</code> on the R command line.

Display Parameter	Description
aggregateGroups	Logical scalar. Aggregate the values within a sample group using the aggre-
assicsate of outpo	gation function specified in the aggregate parameter.
aggregation	Function or character scalar. Used to aggregate values in windows or for
4881.08411011	collapsing overlapping items. The function has to accept a numeric vector
	as a single input parameter and has to return a numeric scalar with the
	aggregated value. Alternatively, one of the predefined options mean, median
	sum, min, max or extreme can be supplied as a character scalar. Defaults to
	mean.
alpha	Numeric scalar between 0 and 1. The opacity of the plotting elements, if
1	supported by the device.
amount	Numeric scalar. Amount of jittering in xy-type plots. See panel.xyplot for
	details.
baseline	Numeric scalar. Y-axis position of an optional baseline. This parameter
	has a special meaning for mountain-type and polygon-type plots, see the
	'Details' section in <i>DataTrack</i> for more information.
box.ratio	Numeric scalar. Parameter controlling the boxplot appearance. See
	panel.bwplot for details.
box.width	Numeric scalar. Parameter controlling the boxplot appearance. See
	panel.bwplot for details.
cex	Numeric scalar. The default pixel size for plotting symbols.
cex.legend	Numeric scalar. The size factor for the legend text.
cex.sampleNames	Numeric scalar. The size factor for the sample names text in heatmap plots.
¢	Defaults to an automatic setting.
coef	Numeric scalar. Parameter controlling the boxplot appearance. See
col	panel.bwplot for details.  Character vector. The base colors to use for all plot types. Unless groups
COI	are specified, only the first color in the vector is usually taken.
col	Character or integer scalar. The color used for all line and symbol elements,
COI	unless there is a more specific control defined elsewhere.
col.baseline	Character scalar. Color for the optional baseline, defaults to the setting of
	col.
col.grid	Integer scalar. The line color for grid elements.
col.histogram	Character scalar. Line color in histogram-type plots.
9	~ · · ·

col.horizon The line color for the segments in the horizon-type plot. See horizonplot for details.

col.line Character or integer scalar. The color used for line elements. Defaults to the setting of col.

col.mountain Character scalar. Line color in mountain-type and polygon-type plots, defaults to the setting of col.

col.sampleNames Character or integer scalar. The color used for the sample names in heatmap plots.

col.symbol Character or integer scalar. The color used for symbol elements. Defaults to the setting of col.

collapse Logical scalar. Collapse overlapping ranges and aggregate the underlying data.

degree Numeric scalar. Parameter controlling the loess calculation for smooth and mountain-type plots. See panel.loess for details.

do.out Logical scalar. Parameter controlling the boxplot appearance. See panel.bwplot for details.

evaluation Numeric scalar. Parameter controlling the loess calculation for smooth and mountain-type plots. See panel.loess for details.

factor Numeric scalar. Factor to control amount of jittering in xy-type plots. See panel.xyplot for details.

family Character scalar. Parameter controlling the loess calculation for smooth and mountain-type plots. See panel.loess for details.

fill Character scalar. The fill color for area elements, unless there is a more specific control defined elsewhere.

fill.histogram Character scalar. Fill color in histogram-type plots, defaults to the setting of fill.

fill.horizon The fill colors for the segments in the horizon-type plot. this should be a vector of length six, where the first three entries are the colors for positive changes, and the latter three entries are the colors for negative changes. Defaults to a red-blue color scheme. See horizonplot for details.

fill.mountain Character vector of length 2. Fill color in mountain-type and polygon-type plots.

fontcolor.legend Integer or character scalar. The font color for the legend text.

fontface.legend Integer or character scalar. The font face for the legend text.

fontfamily.legend Integer or character scalar. The font family for the legend text.

fontsize.legend Numeric scalar. The pixel size for the legend text.

gradient Character vector. The base colors for the 'gradient' plotting type or the 'heatmap' type with a single group. When plotting heatmaps with more than one group, the 'col' parameter can be used to control the group color scheme, however the gradient will always be from white to 'col' and thus does not offer as much flexibility as this 'gradient' parameter.

groups Vector coercable to a factor. Optional sample grouping. See 'Details' section in DataTrack for further information.

h Integer scalar. Parameter controlling the number of vertical grid lines, see panel.grid for details.

horizon.origin The baseline relative to which changes are indicated on the horizon-type plot. See horizonplot for details.

horizon.scale The scale for each of the segments in the horizon-type plot. Defaults to 1/3 of the absolute data range. See horizonplot for details.

jitter.x Logical scalar. Toggle on jittering on the x axis in xy-type plots. See panel.xyplot for details.

jitter.y Logical scalar. Toggle off jittering on the y axis in xy-type plots. See panel.xyplot for details.

legend Boolean triggering the addition of a legend to the track to indicate groups.

This only has an effect if at least two groups are presen.

levels.fos Numeric scalar. Parameter controlling the boxplot appearance. See panel.bwplot for details.

lineheight.legend Numeric scalar. The line height for the legend text.

lty Character or integer scalar. The type for all line elements, unless there is a more specific control defined elsewhere.

lty.baseline Character or numeric scalar. Line type of the optional baseline, defaults to the setting of lty.

lty.grid Integer scalar. The line type for grid elements. Defaults to the setting of lty.

lty.mountain Character or numeric scalar. Line type in mountain-type and polygon-type plots, defaults to the setting of lty.

lwd Integer scalar. The line width for all line elements, unless there is a more specific control defined elsewhere.

lwd.baseline Numeric scalar. Line width of the optional baseline, defaults to the setting of lwd.

lwd.grid Integer scalar. The line width for grid elements. Defaults to the setting of lwd.

lwd.mountain Numeric scalar. Line width in mountain-type and polygon-type plots, defaults to the setting of lwd.

min.distance Numeric scalar. The mimimum distance in pixel below which to collapse ranges.

na.rm Boolean controlling whether to discard all NA values when plotting or to keep empty spaces for NAs

ncolor Integer scalar. The number of colors for the 'gradient' plotting type

notch Logical scalar. Parameter controlling the boxplot appearance. See panel.bwplot for details.

notch.frac Numeric scalar. Parameter controlling the boxplot appearance. See panel.bwplot for details.

pch Integer scalar. The type of glyph used for plotting symbols.

separator Numeric scalar. Number of pixels used to separate individual samples in heatmap-type plots.

showColorBar Boolean. Indicate the data range color mapping in the axis for 'heatmap' or 'gradient' types.

showSampleNames Boolean. Display the names of the individual samples in a heatmap plot.

span Numeric scalar. Parameter controlling the loess calculation for smooth and mountain-type plots. See panel.loess for details.

stackedBars Logical scalar. When there are several data groups, draw the histogram-type plots as stacked barplots or grouped side by side.

stats Function. Parameter controlling the boxplot appearance. See panel.bwplot for details.

transformation Function. Applied to the data matrix prior to plotting or when calling the score method. The function should accept exactly one input argument and its return value needs to be a numeric vector which can be coerced back into a data matrix of identical dimensionality as the input data. Character vector. The plot type, one or several in c("p","1", "b", "a", "s", "g", "r", "S", "smooth", "histogram", "mountain", "polygon", "h", "boxplot", "gradient", "heatmap"). See 'Details' section in DataTrack for more information on the individual plotting types. Integer scalar. Parameter controlling the number of vertical grid lines, see panel.grid for details. varwidth Logical scalar. Parameter controlling the boxplot appearance. See panel.bwplot for details. Numeric or character scalar. Aggregate the rows values of the data matrix window to window equally sized slices on the data range using the method defined in aggregation. If negative, apply a running window of size windowSize using the same aggregation method. Alternatively, the special value auto causes the function to determine the optimal window size to avoid overplotting. windowSize Numeric scalar. The size of the running window when the value of window is negative. Numeric vector of length 2. The range of the y-axis scale. vlim

#### 4.4 AnnotationTrack

Annotation Track objects are the multi-purpose tracks in the Gviz package. Essentially they consist of one or several genomic ranges that can be grouped into composite annotation elements if needed. In principle this would be enough to represent everything from CpG islands to complex gene models, however for the latter the package defines the specialized GeneRegionTrack class, which will be highlighted in a separate section. Most of the features discussed here will also apply to GeneRegionTrack objects, though. As a matter of fact, the GeneRegionTrack class inherits directly from class AnnotationTrack.

AnnotationTrack objects are easily instantiated using the constructor function of the same name. The necessary building blocks are the range coordinates, a chromosome and a genome identifier. Again we try to be flexible in the way this information can be passed to the function, either in the form of separate function arguments, as IRanges or GRanges objects. Optionally, we can pass in the strand information for the annotation features and some useful identifiers. A somewhat special case is to build the object from a GRangesList object, which will automatically preserve the element grouping information contained in the list structure. For the full details on the constructor function and the accepted arguments see ?AnnotationTrack.

```
> aTrack <- AnnotationTrack(start = c(10, 40, 120),
+ width = 15, chromosome = "chrX", strand = c("+",
+ "*", "-"), id = c("Huey", "Dewey", "Louie"),
+ genome = "hg19", name = "foo")
> plotTracks(aTrack)
```

foo

The ranges are plotted as simple boxes if no strand information is available, or as arrows to indicate their direction. We can change the range item shapes by setting the *shape* display parameter. It can also be helpful

to add the names for the individual features to the plot. This can be archived by setting the *showFeatureId* parameter to TRUE

Bulling Dewey Louie

> plotTracks(aTrack, shape = "ellipse", showFeatureId = TRUE,
+ fontcolor = "darkblue")

Bewey Dewey Louie

In this very simplistic example each annotation feature consisted of a single range. In real life the genomic annotation features that we encounter often consists of several sub-units. We can create such composite *AnnotationTrack* objects by providing a grouping factor to the constructor. It needs to be of similar length as the total number of atomic features in the track, i.e, the number of genomic ranges that are passed to the constructor. The levels of the this factor will be used as internal identifiers for the individual composite feature groups, and we can toggle on their printing by setting *showId* to TRUE.

> aTrack.groups <- AnnotationTrack(start = c(50, 180,
+ 260, 460, 860, 1240), width = c(15, 20, 40, 100,
+ 200, 20), chromosome = "chrX", strand = rep(c("+",
+ "\*", "-"), c(1, 3, 2)), group = rep(c("Huey",
+ "Dewey", "Louie"), c(1, 3, 2)), genome = "hg19",
+ name = "foo")
> plotTracks(aTrack.groups, showId = TRUE)

> plotTracks(aTrack, shape = "box", showFeatureId = TRUE)

Buey Dewey Louie

Arranging items on the plotting canvas is relatively straight forward as long as there are no overlaps between invidiual regions or groups of regions. Those inevitably cause overplotting which could seriously obfuscate the information on the plot. A logical solution to this problem is to stack overlapping items in separate horizontal lines to accommodate all of them. This involves some optimization, and the Gviz package automatically tries to come up with the most compact arrangement. Let's exemplify this feature with a slightly modified *AnnotationTrack* object.

> aTrack.stacked <- AnnotationTrack(start = c(50, 180, + 260, 800, 600, 1240), width = c(15, 20, 40, 100, + 500, 20), chromosome = "chrX", strand = "\*", + group = rep(c("Huey", "Dewey", "Louie"), c(1, + 3, 2)), genome = "hg19", name = "foo") > plotTracks(aTrack.stacked, showId = TRUE)

Huey Dewey Louie

We now have our three annotation feature groups distributed over two horizontal lines. One can control the stacking of overlapping items using the *stacking* display parameter. Currently the three values squish, dense and hide are supported. Horizontal stacking is enabled via the squish option, which also is the default. dense forces overlapping items to be joined in one meta-item and hide all together disables the plotting of *AnnotationTrack* items. Please note that adding identifiers to the plot only works for the squish option.

> plotTracks(aTrack.stacked, stacking = "dense")



In addition to annotation groups there is also the notion of a feature type in the Gviz package. Feature types are simply different types of annotation regions (e.g., mRNA transcripts, miRNAs, rRNAs, etc.) that are indicated by different colors. There is no limit on the number of different features, however each element in a grouped annotation item needs to be of the same feature type. We can query and set features using the feature and feature<- methods.

- > feature(aTrack.stacked)
- [1] "unknown" "unknown" "unknown" "unknown" "unknown"
- > feature(aTrack.stacked)[1:4] <- c("foo", "bar", "bar",</pre>
- + "bar")

Huey

Unless we tell the Gviz package how to deal with the respective feature types they will all be treated in a similar fashion, i.e., they will be plotted using the default color as defined by the *fill* display parameter. To define colors for individual feature types we simply have to add them as additional display parameters, where the parameter name matches to the feature type and its value is supposed to be a valid R color qualifier. Of course this implies that we can only use type names that are not already taken by other display parameters defined in the package.

```
> plotTracks(aTrack.stacked, showId = TRUE, foo = "darkred",
+ bar = "darkgreen")
```

Stacking of annotation items to avoid overplotting only works as long as there is enough real estate on the plotting canvas to separate all items, i.e., we need all items to be at least a single pixel wide to correctly display them. This limitation is automatically enforced by the Gviz package, however it implies that unless neighbouring items are more than one pixel appart we can not distinguish between them and will inevitably introduce a certain amount of overplotting. This means that on a common screen device we can only look at a very limited genomic region of a few kb in full resolution. Given that an average chromosome is in the order of a few gb, we still need a reasonable way to deal with the overplotting problem despite the item stacking functionality. As default, the Gviz package will merge all overlapping items into one unified meta-item and only plot that (see 'Collapse' section below for details). In order to indicate the amount of overplotting that was introduced by this process we can use the *showOverplotting* display parameter. It uses a color scale (based on the original colors defined for the track), with lighter colors indicating areas of low or no overplotting, and more saturated colors indicating areas of high overplotting density. We exemplify this feature on an *AnnotationTrack* object that represents a good portion of a real human chromosome.

Louie

- > data("denseAnnTrack")
- > plotTracks(denseAnnTrack, showOverplotting = TRUE)



#### Collapsing

All track types that inherit from class AnnotationTrack support the collapsing of overlapping track items, either because they have initially been defined as overlapping coordinates, or because the current device resolution does not allow to sufficiently separate them. For instance, two elements of a feature group may be separated by 100 base pairs on the genomic scale, however when plotted to the screen, those 100 base pairs translate to a distance of less than one pixel. In this case we can no longer show the items as two separate entitites. One solution to this problem would be to allow for arbitrary overplotting, in which case the last one of the overlapping items that is drawn on the device wins. This is not optimal in many ways, and it also poses a significant burden on the graphical engine because a lot of stuff has to be drawn which no one will ever see.

To this end the Gviz package provides an infrastructure to reasonably collapse overlappig items, thereby adjusting the information content that can be shown to the available device resolution. By default this feature is turned on, and the user does not have to worry too much about it. However, one should be aware of the consequences this may have on a given visualization. If you absolutely do not want collapsing to take place, you may completely turn it off by setting the display parameter collapse to FALSE. Please note that by doing this the showOverplotting parameter will also stop working. If you opt in, there is some considerable amount of detailed control to fine tune the collapsing to your needs.

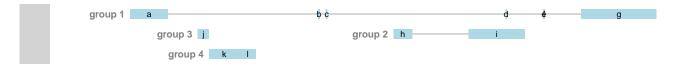
Lets start with a small example track for which element collapsing has been turned off and no adjustments to the ranges have been made. We plot both the item identifiers and the group identifiers to exemplify what is going on.

- > data(collapseTrack)
- > plotTracks(ctrack, extend.left = 1800)



The first thing to notice is that the for item d we do see the item identifier but not the range itself. This is due to the fact that the with of the item is smaller than a single pixel, and hence the graphics system can not display it (Note that this is only true for certain devices. The quartz device on the Mac seems to be a little smarter about this). There are also the two items e and f which seem to overlay each other completely, and another two items which appear to be one joined item (k and 1). Again, this is a resolution issue as their relative distance is smaller than a single pixel, so all we see is a single range and some ugly overplotted identifiers. We can control the first issue by setting the minimum pixel width of a plotted item to be one pixel using the min.width display parameter.

> plotTracks(ctrack, extend.left = 1800, min.width = 1)



Now the item d has a plotable size and can be drawn to the device. The overplotted items are still rather anoying, but the only way to get rid of those is to turn item collapsing back on.

Now all items that could not be separated by at least one pixel have been merged into a single meta-item, and the software has taken care of the identifiers for you, too. The merging operation is aware of the grouping information, so no two groups where joint together. Sometimes a single pixel width or a single pixel distance is not enough to get a good visualization. In these cases one could decide to enforce even larger values. We can do this not only for the minimum width, but also for the minimum distance by setting the min.distance parameter.

This time also the two items b and c have been merged, and all ranges are now at least 3 pixels wide. Depending on the density of items on the plot even this reduction can be insufficient. Because we did not merge complete groups we might still end up with quite a lot of stacks to accommodate all the information. To this end the display parameter mergeGroups can be used to disable absolute group separation. Rather than blindly merging all groups (as it is done when stacking='dense') however, the software will only join those overlapping group ranges for which all items are already merged into a single meta item.

#### Building AnnotationTrack objects from files

Just like we showed before for *DataTrack* objects it is possible to stream the data for an *AnnotationTrack* object off the file system. We will again use the bam file as an example. This time the default import function reads the coordinates of all the sequence alignments from the file and even knows how to deal with grouped elements.

Since AnnotationTrack objects are somewhat more complicated than DataTrack objects, the constructor provides another level of flexibility when reading the data from a file. More specifically, the user has the ability to map the provided elementMetadata columns of the GRanges object that is returned by the import function to the respective columns in the final AnnotationTrack object. If no explicit mapping is provided, the package will try to find a default mapping based on the input file type and the desired track type. The current mapping of a track is displayed by its show method as can be seen above. The mechanism for changing this default mapping is quite straight forward: if the values in the relevant constructor arguments (in this case group, id and feature) are character scalars, the elementMetadata columns of the GRanges object that is returned by the import function are mapped against their values. For instance we can map the id column to the group attribute and thus suppress the grouping of sequence alignments that originate from the same sequencing read.



Please note that if no *elementMetadata* can be mapped to the constructor arguments their values are assumed to be normal character scalars and they will be recycled accordingly to fill the track's slots. One can ask for the package's default mappings by using the availableDefaultMapping function.

# > availableDefaultMapping(bamFile, "AnnotationTrack")

\$id [1] "id"

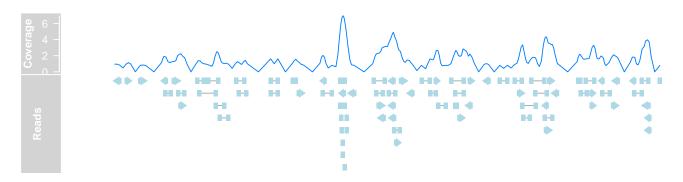
# \$group

[1] "group"

We can now plot both the *DataTrack* representation as well as the AnnotationTrack representation of the bam file together to prove that the underlying data are indeed identical.

> plotTracks(list(dTrack4, aTrack2), from = 189990000,

+ to = 190000000)



#### Display parameters for AnnotationTrack objects

For a complete listing of all the available display parameters please see the table below or the man page of the *AnnotationTrack* class by typing in ?AnnotationTrack on the R command line.

Display Parameter	Description
alpha	Numeric scalar between 0 and 1. The opacity of the plotting elements, if
	supported by the device.
cex	Numeric scalar. The font expansion factor for item identifiers.
cex.group	Numeric scalar. The font expansion factor for the group-level annotation.
col	Character or integer scalar. The border color for all track items.
col.line	Character scalar. The color used for connecting lines between grouped items.
	Defaults to a light gray, but if set to NULL the same color as for the first item
	in the group is used.
fill	Character or integer scalar. The fill color for untyped items. This is also
	used to connect grouped items. See grouping for details.
fontcolor	Character or integer scalar. The font color for item identifiers.
fontcolor.group	Character or integer scalar. The font color for the group-level annotation.
fontface	Integer scalar. The font face for item identifiers.
fontface.group	Numeric scalar. The font face for the group-level annotation.
fontface.group	Numeric scalar. The font face for the group-level annotation.

fontfamily Character scalar. The font family for item identifiers. fontfamily.group Character scalar. The font family for the group-level annotation. fontsize Numeric scalar. The font size for item identifiers. fontsize.group Numeric scalar. The font size for the group-level annotation. Numeric scalar. The line expansion factor for all track items. This is also used to connect grouped items. See grouping for details. lineheight Numeric scalar. The font line height for item identifiers. Character or integer scalar. The line type for all track items. This is also used to connect grouped items. See grouping for details. Integer scalar. The line width for all track items. This is also used to connect lwd grouped items. See grouping for details. mergeGroups Logical scalar. Merge fully overlapping groups if collapse==TRUE. min.height Numeric scalar. The minimum range height in pixels to display. All ranges are expanded to this size in order to avoid rendering issues. See collapsing for details. For feathered bars indicating the strandedness of grouped items this also controls the height of the arrow feathers. Numeric scalar. The minimum range width in pixels to display. All ranges min.width are expanded to this size in order to avoid rendering issues. See collapsing for details. rotation Numeric scalar. The degree of text rotation for item identifiers. shape Character scalar. The shape in which to display the track items. Currently only box, arrow, ellipse, and smallArrow are implemented. showFeatureId Logical scalar. Control whether to plot the individual track item identifiers. showId Logical scalar. Control whether to annotate individual groups. showOverplotting Logical scalar. Use a color gradient to show the amount of overplotting for collapsed items. This implies that collapse==TRUE

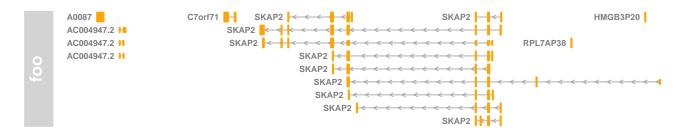
#### 4.5 GeneRegionTrack

GeneRegionTrack objects are in principle very similar to AnnotationTrack objects. The only difference is that they are a little more gene/transcript centric, both in terms of plotting layout and user interaction, and that they may define a global start and end position (a feature which is not particularly relevant for the normal user). The constructor function of the same name is a convenient tool to instantiate the object from a variety of different sources. In a nutshell, we need to pass start and end positions (or the width) of each annotation feature in the track and also supply the exon, transcript and gene identifiers for each item which will be used to create the transcript groupings. A somewhat special case is to build a GeneRegionTrack object directly from one of the popular TranscriptDb objects, an option that is treated in more detail below. For more information about all the available options see the class's manual page (?GeneRegionTrack).

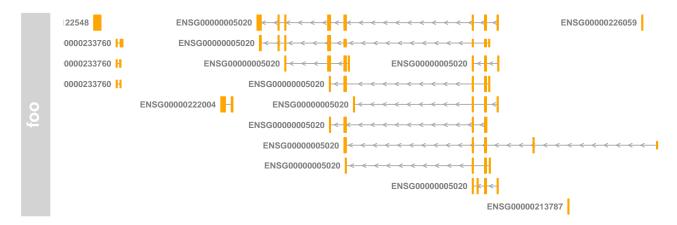
There are a number of accessor methods that make it easy to query and replace for instance exon, transcript or gene assignments. There is also some support for gene aliases or gene symbols which are often times more useful than cryptic data base gene identifiers. The following code that re-uses the GeneRegionTrack object from the first section exemplifies some of these features.

```
> data(geneModels)
> grtrack <- GeneRegionTrack(geneModels, genome = gen,
+ chromosome = chr, name = "foo")
> head(gene(grtrack))
> head(transcript(grtrack))
```

- > head(exon(grtrack))
- > head(symbol(grtrack))
- > plotTracks(grtrack, extend.left = 20000, showId = TRUE)



- > plotTracks(grtrack, extend.left = 20000, showId = TRUE,
- geneSymbols = FALSE)



Since we have the gene level information as part of our GeneRegionTrack objects we can ask the package to collapse all of our gene models from individual exons and transcripts down to gene body locations by setting the collapse Transcripts display parameter to TRUE.

```
> plotTracks(grtrack, collapseTranscripts = TRUE, shape = "arrow",
```

showId = TRUE, extend.left = 20000)



#### Building GeneRegionTrack objects from TranscriptDbs

The GenomicFeatures packages provides an elegant framework to download gene model information from online sources and to store it locally in a SQLite data base. Because these so called TranscriptDb objects have become the de-facto standard for genome annotation information in Bioconductor we tried to make it as simple as possible to convert them into GeneRegionTracks. Essentially one only has to call the constructor function with the TranscriptDb object as a single argument. We exemplify this on a small sample data set that comes with the GenomicFeatures package.

```
> library(GenomicFeatures)
> samplefile <- system.file("extdata", "UCSC_knownGene_sample.sqlite",
      package = "GenomicFeatures")
> txdb <- loadDb(samplefile)</pre>
> GeneRegionTrack(txdb)
GeneRegionTrack 'GeneRegionTrack'
| genome: hg18
| active chromosome: chr1
| annotation features: 9
There are 852 additional annotation features on 45 further chromosomes
  chr10: 23
  chr11: 9
  chr12: 31
  chr13: 3
  chr13_random: 11
  chr9_random: 16
  chrM: 3
  chrX: 21
  chrX_random: 16
  chrY: 21
Call seglevels(obj) to list all available chromosomes or seginfo(obj) for more detailed output
```

In this context, the constructor's *chromosome*, *start* and *end* argument take on a slightly differnt meaning in that they can be used to subset the data that is fetched from the *TranscriptDb* object. Please note that while the *chromosome* alone can be supplied, providing *start* or *end* without the chromosome information will not work.

```
> txTr <- GeneRegionTrack(txdb, chromosome = "chr6",
+ start = 300000, end = 350000)</pre>
```

Call chromosome(obj) <- 'chrId' to change the active chromosome

A nice bonus when building GeneRegionTracks from TranscriptDb objects is that we get additional information about coding and non-coding regions of the transcripts, i.e., coordinates of the 5' and 3' UTRs and of the CDS regions. The class' plotting method can use this inforantion to distinguish between coding and non-coding regions based on the shape of the feature. All coding regions are plotted just as we have seen in the previous examples, whereas the non-coding regions are drawn as slighly thinner boxes. The distinction between coding and non-coding is made on the basis of the object's feature values in combination with a special display parameter thinBoxFeature that enumerates all feature types that are to be treated as non-coding. Obviously this feature is available to all GeneRegionTracks, not only the ones that were build from TranscriptDb objects. However, the coding information has to be added manually and the default value of the thinBoxFeature parameter may not be sufficient to cover all possible cases. It is up to the user to come up with a complete list of non-coding feature types depending on the source of the data.

```
> feature(txTr)
> plotTracks(txTr, showId = TRUE, extend.left = 1000)
```

# ${\bf Display\ parameters\ for\ GeneRegionTrack\ objects}$

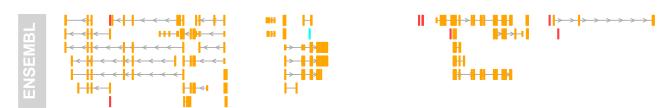
For a complete listing of all the available display parameters please see the table below or the man page of the GeneRegionTrack class by typing in <code>?GeneRegionTrack</code> on the R command line.

Display Parameter	Description					
alpha	Numeric scalar between 0 and 1. The opacity of the plotting elements, is					
-	supported by the device.					
cex	Numeric scalar. The font expansion factor for item identifiers.					
cex.group	Numeric scalar. The font expansion factor for the group-level annotation.					
col	Character or integer scalar. The border color for all track items. Defaults to using the same color as in fill, also taking into account different track features.					
${\bf collapse Transcripts}$	Logical scalar. Merge all transcripts of the same gene into one single ge model. Essentially, this will only keep the start location of the first exon at the end location of the last exon from all transcripts of a gene.					
fill	Character or integer scalar. The fill color for untyped items. This is also used to connect grouped items. See grouping for details.					
fontcolor	Character or integer scalar. The font color for item identifiers.					
fontcolor.group	Character or integer scalar. The font color for the group-level annotation.					
fontface	Integer scalar. The font face for item identifiers.					
fontface.group	Numeric scalar. The font face for the group-level annotation.					
fontfamily	Character scalar. The font family for item identifiers.					
fontfamily.group	Character scalar. The font family for the group-level annotation.					
fontsize	Numeric scalar. The font size for item identifiers.					
fontsize.group	Numeric scalar. The font size for the group-level annotation.					
geneSymbols	Logical scalar. Use human-readable gene symbols or gene IDs for the transcript annotation.					
lex	Numeric scalar. The line expansion factor for all track items. This is also used to connect grouped items. See grouping for details.					
lineheight	Numeric scalar. The font line height for item identifiers.					
lty	Character or integer scalar. The line type for all track items. This is also used to connect grouped items. See grouping for details.					
lwd	Integer scalar. The line width for all track items. This is also used to connect grouped items. See grouping for details.					
min.distance	Numeric scalar. The minimum pixel distance before collapsing range items, only if collapse==TRUE. See collapsing for details. Note that a value larger than 0 may lead to UTR regions being merged to CDS regions, which in most cases is not particularly useful.					
min.width	Numeric scalar. The minimum range width in pixels to display. All ranges are expanded to this size in order to avoid rendering issues. See collapsing for details.					
rotation	Numeric scalar. The degree of text rotation for item identifiers.					
shape	Character scalar. The shape in which to display the track items. Currently only box, arrow, ellipse, and smallArrow are implemented.					
${\rm show} {\rm Exon} {\rm Id}$	Logical scalar. Control whether to plot the individual exon identifiers.					
showId	Logical scalar. Control whether to annotate individual groups.					
showOverplotting	Logical scalar. Use a color gradient to show the amount of overplotting for collapsed items. This implies that collapse==TRUE					

### 4.6 BiomartGeneRegionTrack

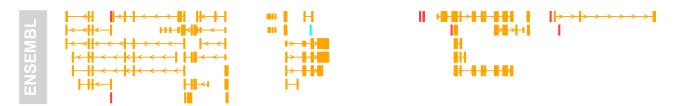
As seen before it can be very useful to quickly download gene annotation information from an online repositry rather than having to construct it each time from scratch. To this end, the Gviz package also defines the BiomartGeneRegionTrack class, which directly extends GeneRegionTrack but provides a direct interface to the ENSEMBL Biomart service (yet another interface to the UCSC data base content is highlighted in one of the next sections). Rather than providing all the bits and pieces for the full gene model, we just enter a genome, chromosome and a start and end position on this chromosome, and the constructor function BiomartGeneRegionTrack will automatically contact ENSEMBL, fetch the necessary information and build the gene model on the fly. Please note that you will need an internet connection for this to work, and that contacting Biomart can take a significant amount of time depending on usage and network traffic. Hence the results are almost never going to be returned instantaniously.

- > biomTrack <- BiomartGeneRegionTrack(genome = "hg19",</pre>
- + chromosome = chr, start = 20000000, end = 21000000,
- + name = "ENSEMBL")
- > plotTracks(biomTrack)



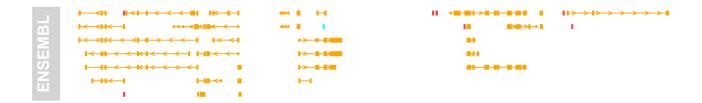
You may have noticed in the above plot that the track includes feature information which is displayed by the different feature colors and box sizes. This information has been automatically extracted from Biomart and the respective color coding is part of the class' definition. We can highlight the feature classes even more by using similarly colored bars to connect the grouped elements.

> plotTracks(biomTrack, col.line = NULL)



For dense stacked plots like this it can sometimes be helpful to control the vertical spacing between the stacking rows. A simple way to do this is available for each object inheriting form the *StackedTrack* class by means of the *stackHeight* display parameter. It takes a value between 0 and 1 that controls which fraction of the available vertical space on a stacking line should be used to draw the feature glyphs.

> plotTracks(biomTrack, col.line = NULL, stackHeight = 0.3)



#### Display parameters for BiomartGeneRegionTrack objects

For a complete listing of all the available display parameters please see the table above in the previous GeneRegionTrack section or the man page of the BiomartGeneRegionTrack class by typing in ?BiomartGeneRegionTrack on the R command line.

As mentioned above, one additional benefit when fetching the data through Biomart is that we also receive some information about the annotation feature types, which is automatically used for the color coding of the track. The following table shows the available feature types.

Display Parameter	Description	Color
C_segment	Character or integer scalar. Fill color for annotation objects of type 'C_segment'.	burlywood4
D_segment	Character or integer scalar. Fill color for annotation objects of type 'C_segment'.	lightblue
J_segment	Character or integer scalar. Fill color for annotation objects of type 'C_segment'.	dodgerblue2
$Mt_rRNA$	Character or integer scalar. Fill color for annotation objects of type 'Mt_rRNA'.	yellow
$Mt_tRNA$	Character or integer scalar. Fill color for annotation objects of type 'Mt_tRNA'.	darkgoldenrod
$Mt_{tRNA_pseudogene}$	Character or integer scalar. Fill color for annotation objects of type 'Mt_tRNA_pseudogene'.	${\rm darkgolden} {\rm rod} 1$
$V_{\rm segment}$	Character or integer scalar. Fill color for annotation objects of type 'V_segment'.	aquamarine
miRNA	Character or integer scalar. Fill color for annotation objects of type 'L_segment'.	cornflowerblue
$miRNA\_pseudogene$	Character or integer scalar. Fill color for annotation objects of type 'miRNA_pseudogene'.	cornsilk
$misc\_RNA$	Character or integer scalar. Fill color for annotation objects of type 'misc_RNA'.	cornsilk3
$misc\_RNA\_pseudogene$	Character or integer scalar. Fill color for annotation objects of type 'misc_RNA_pseudogene'.	cornsilk4
protein_coding	Character or integer scalar. Fill color for annotation objects of type 'protein_coding'.	orange
pseudogene	Character or integer scalar. Fill color for annotation objects of type 'pseudogene'.	brown1
rRNA	Character or integer scalar. Fill color for annotation objects of type 'rRNA'.	darkolivegreen1
rRNA_pseudogene	Character or integer scalar. Fill color for annotation objects of type 'rRNA_pseudogene'.	darkolivegreen

retrotransposed	Character or integer scalar. Fill color for annotation objects of type 'retrotransposed'.	blueviolet
scRNA	Character or integer scalar. Fill color for annotation objects of type 'scRNA'.	gold4
scRNA_pseudogene	Character or integer scalar. Fill color for annotation objects of type 'scRNA_pseudogene'.	darkorange2
$\operatorname{snRNA}$	Character or integer scalar. Fill color for annotation objects of type 'snRNA'.	coral
$snRNA\_pseudogene$	Character or integer scalar. Fill color for annotation objects of type 'snRNA_pseudogene'.	coral3
snoRNA	Character or integer scalar. Fill color for annotation objects of type 'snoRNA'.	cyan
$snoRNA\_pseudogene$	Character or integer scalar. Fill color for annotation objects of type 'snoRNA_pseudogene'.	cyan2
tRNA_pseudogene	Character or integer scalar. Fill color for annotation objects of type 'tRNA_pseudogene'.	antiquewhite3

#### 4.7 DetailsAnnotationTrack

It is sometimes desirable to add more detailed information to particular ranges in an Gviz plot for which the notion of genomic coordinates no longer makes sense. For instance, the ranges in an AnnotationTrack may represent probe locations on a genome, and for each of these probes a number of measurements from multiple samples and from different sample groups are available. To this end, the DetailsAnnotationTrack provides a flexible interface to further annotate genomic regions with arbitrary additional information. This is archived by splitting the AnnotationTrack plotting region into two horizontal sections: the lower section containing the range data in genomic coordinates, and the upper one containing the additional data for each of the displayed ranges in vertically tiled panels of equal size. The connection between a range item and its details panel is indicated by connecting lines.

The content of the individual details panels has to be filled in by a user-defined plotting function that uses grid (or lattice) plotting commands. This function has to accept a number of mandatory parameters, notably the start, end, strand, chromosome and identifier information for the genomic range, as well as an integer counter indicating the index of the currently plotted details tile. This information can be used to fetch abtritray details, e.g. from a list, and environement or even from a *GRanges* object which will then be processed and visualized within the plotting function. This may sound rather abstract, and for more details please refer to the class' help page. For now we just want to demonstrate the functionality in a simple little example. We begin by defining a *GRanges* object containing 4 genomic locations. In our example those are considered to be probe locations from a methylation array.

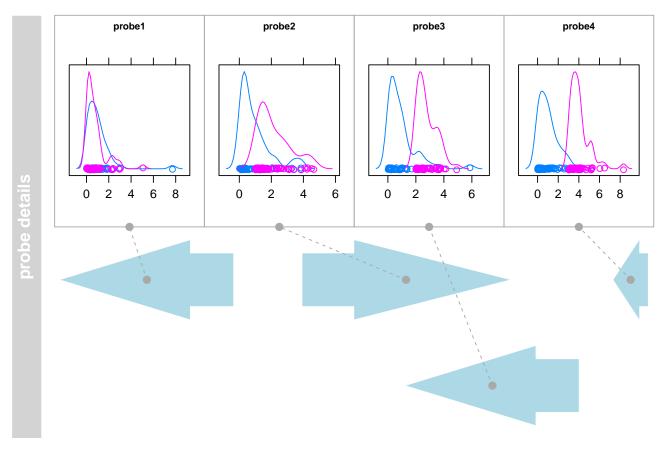
```
> library(GenomicRanges)
> probes <- GRanges(seqnames = "chr7", ranges = IRanges(start = c(2000000,
+ 2070000, 2100000, 2160000), end = c(2050000,
+ 2130000, 2150000, 2170000)), strand = c("-",
+ "+", "-", "-"))</pre>
```

For each of these probes we have methylation measurements from a large number of different samples in a numeric matrix, and within the samples there are two treatment groups. The aim is to compare the distribution of measurement values between these two groups at each probe locus.

Of course we could use a *DataTrack* with the box-plot representation for this task, however we do have strand-specific data here and some of the probes can be overlapping, so all this information would be lost. We are also interested in the particular shape of the data distribution, so a density plot representation is what we really need. Luckily, the lattice package gives us a nice densityplot function that supports grouping, so all that's left to do now is to write a little wrapper that handles the extraction of the relevant data from the matrix. This is easily archieved by using the range identifiers, which conveniently map to the row names of the data matrix.

Finaly, it is as simple as calling the *AnnotationTrack* constructor, passing along the wrapper function and calling plotTracks.

```
> deTrack <- AnnotationTrack(range = probes, genome = "hg19",
+ chromosome = 7, id = rownames(methylation), name = "probe details",
+ stacking = "squish", fun = details)
> plotTracks(deTrack)
```



It should be noted here that in our little example we rely on the methylation data matrix and the grouping vector to be present in the working environment. This is not necessarily the cleanest solution and one should consider storing additional data in an evironment, passing it along using the *detailFunArgs* parameter, or making it part of the details function in form of a closure. The class' help page provides further instructions.

Another use case for the *DetailsAnnotationTrack* class is to deal with the problem of very different feature sizes within a single track. For instance, we may be looking at a rather large genomic region containing one big transcript with many widely spaced exons and a bunch of smaller, more compact transcripts. In this case it would be helpful to provide a zoomed in version of those smaller transcripts. In order to achieve this we can make use of the class' groupDetails display parameter, which applies the detail plotting function over each range group rather than over individual range items. First we define a function that selects those groups with a plotted size smaller than 10 pixels. We make use of the unexported function .pxResolution here to come up with the mapping between pixel coordinates and genomic coordinates.

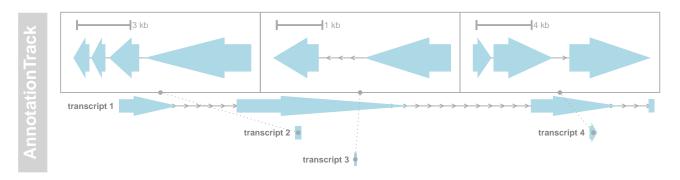
The actual detail plotting function is fairly trivial. In the details viewport we simply call plotTracks function on the subset of items from each group that has been selected before without plotting the track

titles, and also add a scale indicator for some reference. It is worth mentioning however that we need to extract the ranges of the zoomed in track items from the unmodified track object (the Gdobject.original argument in the details function) because the object that is plotted in the lower panel contains only the collapsed ranges.

```
> detFun <- function(identifier, GdObject.original,
+ ...) {
+ plotTracks(list(GenomeAxisTrack(scale = 0.3,
+ size = 0.2, cex = 0.7), GdObject.original[group(GdObject.original) ==
+ identifier]), add = TRUE, showTitle = FALSE)
+ }</pre>
```

Finally, we load some sample data, turn it into a DetailsAnnotationTrack object and plot it.

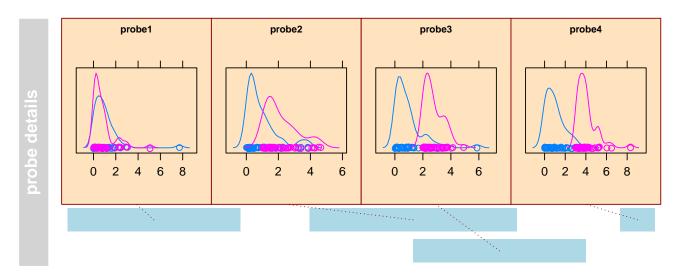
```
> data(geneDetails)
> deTrack2 <- AnnotationTrack(geneDetails, fun = detFun,
+ selectFun = selFun, groupDetails = TRUE, details.size = 0.5,
+ detailsConnector.cex = 0.5, detailsConnector.lty = "dotted",
+ shape = c("smallArrow", "arrow"), showId = TRUE)
> plotTracks(deTrack2, extend.left = 90000)
```



#### Display parameters for DetailsAnnotationTrack objects

In addition to the display parameters for the *AnnotationTrack* class, some additional parameters can be used to control the look and feel of the details sections. For a complete listing of all the available display parameters please see the tables below and the one above in the *AnnotationTrack* section or the man page of the *DetailsAnnotationTrack* class by typing in ?DetailsAnnotationTrack on the R command line.

```
> plotTracks(deTrack, details.size = 0.75, detailsConnector.pch = NA,
+ detailsConnector.col = "darkred", detailsBorder.fill = "#FFE3BF",
+ detailsBorder.col = "darkred", shape = "box",
+ detailsConnector.lty = "dotted")
```



Display Parameter	Description				
details.minWidth	Numeric scalar. The minium width in pixels for a details panel, if less space				
details.iiiii widtii	is available no details are plotted.				
1	*				
details.ratio	Numeric scalar. By default, the plotting method tries to fill all available				
	space of the details panel tiles. Depending on the dimensions of your plot				
	and the number of tiles this may lead to fairly stretched plots. Restricting				
	the ration of width over height can help to fine tune for somewhat more				
	sane graphics in these cases. Essentially this adds some white space in be-				
	tween individual tiles to force the desired ratio. Together with the size and				
	details.size arguments, which control the vertical extension of the whole				
	track and of the details section, this allows for some fairly generic resizing				
	of the tiles.				
details.size	Numeric scalar. The fraction of vertical space of the track used for the details				
	section.				
detailsBorder.col	Character or integer scalar. Line color of the border.				
detailsBorder.fill	Character or integer scalar. Background color of the border.				
detailsBorder.lty	Character or integer scalar. Line type of the border around each detail				
	panel.				
detailsBorder.lwd	Integer scalar. Line width of the border.				
detailsConnector.cex	Numeric scalar. Relative size of the connector's end points.				
detailsConnector.col	Character or integer scalar. Color of the line connecting the Annotstion-				
	Track item with its details panel.				
detailsConnector.lty	Character or integer scalar. Type of connecting line.				
${\it details} Connector. lwd$	Integer scalar. Line width of the connector.				
detailsConnector.pch	Integer scalar. Type of the connector's ends.				
detailsFunArgs	List.Additional arguments that get passed on the the details plotting func-				
	tion.				
groupDetails	Logial scalar. Plot details for feature groups rather than for individual fea-				
	tures.				

## 4.8 SequenceTrack

So far we have displayed all kinds of ranges on a genome, but never really looked at the underlying genomic sequence. To that end, the Gviz package defines the SequenceTrack class which can hold genomic sequence information derived from either a DNAStringSet object of, even more convenient, from one of the BSgenome packages. There is not formal checking of the track's genome, so it is up to the user to provide reasonable inputs to the class' constructor. As with all the other track types, the constructor has the same name as the class. In this example we build a SequenceTrack from the human hg19 UCSC genome.

```
> library(BSgenome.Hsapiens.UCSC.hg19)
> sTrack <- SequenceTrack(Hsapiens)
> sTrack
Sequence track 'SequenceTrack':
| genome: hg19
| chromosomes: 93
| active chromosome: chr1 (249250621 nulceotides)
Call segnames() to list all available chromosomes
Call chromosome()<- to change the active chromosome
Parent BSgenome object:
| organism: Homo sapiens (Human)
| provider: UCSC
| provider version: hg19
| release date: Feb. 2009
| release name: Genome Reference Consortium GRCh37
| package name: BSgenome.Hsapiens.UCSC.hg19
```

Here we retain the same benefits as for regular *BSgenome* objects in that the actual sequence is only loaded into memory when a particular chromosome is first accessed. Plotting the track yields the expected result: a character representation of the genomic sequence in the current plot window. It may be worth noting that the color scheme that is used to encode the indvidual nucleotides is defined in the biovizBase package and the user is referred to its documentation for more details.

```
> plotTracks(sTrack, chromosome = 1, from = 20000,
+ to = 20050)
```

# T C C T G G T G C T C C C A C A A A G G A G A A G G G C T G A T C A C T C A A A G T T G C G A A C A

Sometimes it may be too busy to plot both the sequence letters and the color coding, and we can assign arbitrary colors by setting the fontcolor display parameter. The convention here is that we need to supply a named vector, with one entry for each of the five possible nucleotides (A, C, T, G, and N). If any of the entries is missing, the respective letter will not be drawn at all.

```
> fcol <- c(A = "darkgray", C = "darkgray", T = "darkgray",
+ G = "darkgray")
> plotTracks(sTrack, chromosome = 1, from = 20000,
+ to = 20050, fontcolor = fcol)
```

#### T C C T G G T G C T C C C A C A A A G G A G A A G G G C T G A T C A C T C A A A G T T G C G A A C A

In this case we are plotting the sequence on the forward strand and we may want to make this clear by adding direction indicators.

```
> plotTracks(sTrack, chromosome = 1, from = 20000,
+ to = 20050, add53 = TRUE)
5' CCTGGTGCTCCCACACAAGGAGAGGGCTGATCACTCAAAGTTGCGAAC3'
```

If instead we plot the complement sequence on the reverse strand, the indicators are automatically adjusted.

```
> plotTracks(sTrack, chromosome = 1, from = 20000,
+ to = 20050, add53 = TRUE, complement = TRUE)
```

#### 3' GGACCACGAGGGTGTTTCCTCTTCCCGACTAGTGAGTTTCAACGCTTTG 5'

So far we have been able to fit the sequence onto our plotting device without overplotting. I.e., all the letters nicely fit in to their respective position in the genomic coordinate system. However this restricts us to very small windows which we can reasonably display. A much more compact version of the same information is to use colored boxes rather than individual letters. Those boxes can be stacked much closer together, which increases the possible window size quite a bit. The user does not really need to worry about this as the plotting method will automatically make a reasonable decision based on the available space.

```
> plotTracks(sTrack, chromosome = 1, from = 20000,
+ to = 20100)
```

For added flexibility one can set the noLetters display parameter to TRUE to always force the drawing of boxes. Of course also the colored box representation of a sequence has its limits, and if we cross the threshold when individual boxes can not be separated anymore, the plotting method falls back to drawing a single line indicating the presence of a sequence at the given position.

```
> plotTracks(sTrack, chromosome = 1, from = 20000,
+ to = 201000)
```

Finally, the selected font size is also contributing to the available space, and we can cram a little more sequence into a given window by decreasing it.

```
> plotTracks(sTrack, chromosome = 1, from = 20000,
+ to = 20100, cex = 0.5)
```

## 4.9 Creating tracks from UCSC data

The UCSC data bases contain a multitude of genome annotation data for dozents of different organisms. Some of those data are very simple annotations like CpG island locations or SNP locations. Others are more complicated gene models, or even numeric annotations like conservation information. In order to provide a unified interface to all this information, the Gviz package defines a meta-constructor function UcscTrack. The idea here is that we can express all of the available Ucsc data in one of the package's track types. We use the functionality provided in the rtracklayer package to connect to UCSC and to download the relevant information. As a little illustrative example, let's reproduce a view from the famous UCSC genome browser using the Gviz package. As a final result we want to show something similar to Figure 4.9.

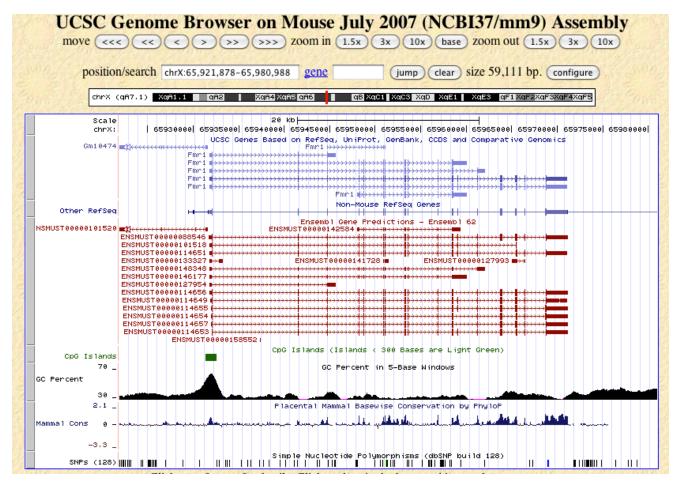


Figure 1: A screen shot of a UCSC genome browser view around the FMR1 locus on the mouse chromosome.

To start we first need to know about the available data in the UCSC data base and about their structure. A good way to do this is to use the table browser on the UCSC web site (http://genome.ucsc.edu/cgi-bin/hgTables?command=start). Figure 4.9 shows the table structure for the first gene model track, the known UCSC genes, in the table browser. We can see that there are multiple fields, some with genomic locations, other with additional data like labels or identifiers. If we go back to the section about the GeneRegionTrack class we remember that we need exactly this type of information for the constructor function. So in order to take the UCSC data and build an object of class GeneRegionTrack we need a way to map them to the individual constructor arguments. This is exactly what the UcscTrack meta-constructor is supposed to do for us.

Format description: Genes based on RefSeq, GenBank, and UniProt.						
field	example	SQL type	info	description		
name	uc007aet.1	varchar(255)	values	Name of gene		
chrom	chr1	varchar(255)	values	Reference sequence chromosome or scaffold		
strand	-	char(1)	values	+ or - for strand		
txStart	3195984	int(10) unsigned	range	Transcription start position		
txEnd	3205713	int(10) unsigned	range	Transcription end position		
cdsStart	3195984	int(10) unsigned	range	Coding region start		
cdsEnd	3195984	int(10) unsigned	range	Coding region end		
exonCount	2	int(10) unsigned	range	Number of exons		
exonStarts	3195984,3203519,	longblob		Exon start positions		
exonEnds	3197398,3205713,	longblob		Exon end positions		
proteinID		varchar(40)	values	UniProt display ID for Known Genes, UniProt accession or RefSeq protein ID for UCSC Ger		
alignID	uc007aet.1	varchar(255)	values	Unique identifier for each (known gene, alignment position) pair		

Figure 2: A screen shot of a UCSC table browser view on the UCSC Known Genes track.

It needs to know about the track for which to extract the data (and optionally one or several of the tables that make up the collective track data, see <code>?UcscTrack</code> for details), about the genomic range including the chromosome for which to extract data, about the type of <code>Gviz</code> track that we want to translate this data into, and about the individual track columns and their counterparts in the respective track class constructor. In our example, the track is called <code>knownGene</code>, the track type to construct is <code>GeneRegionTrack</code>, and the relevant columns are <code>exonStarts</code>, <code>exonEnds</code>, <code>name</code> and <code>strand</code>, which we will use as the start and end coordinates of the ranges and for all the exon, transcript and gene identifiers. Here we make use of the high flexibility of the <code>GeneRegionTrack</code> constructor in the sense that the exon coordinates actually come in the form of a comma-separated list, combining all the information for one transcript in one row of the table. The function is smart enough to detect this and to split the annotation regions accordingly. The full function call to create the <code>GeneRegionTrack</code> from the UCSC data looks like this:

```
> from <- 65921878
> to <- 65980988
> knownGenes <- UcscTrack(genome = "mm9", chromosome = "chrX",
+ track = "knownGene", from = from, to = to, trackType = "GeneRegionTrack",
+ rstarts = "exonStarts", rends = "exonEnds", gene = "name",
+ symbol = "name", transcript = "name", strand = "strand",
+ fill = "#8282d2", name = "UCSC Genes")</pre>
```

With a similar approach we can construct the next two gene model tracks based on the **xenoRefGene** and **ensGene** data tables.

```
> refGenes <- UcscTrack(genome = "mm9", chromosome = "chrX",
+ track = "xenoRefGene", from = from, to = to,
+ trackType = "GeneRegionTrack", rstarts = "exonStarts",
+ rends = "exonEnds", gene = "name", symbol = "name2",
+ transcript = "name", strand = "strand", fill = "#8282d2",
+ stacking = "dense", name = "Other RefSeq")
> ensGenes <- UcscTrack(genome = "mm9", chromosome = "chrX",
+ track = "ensGene", from = from, to = to, trackType = "GeneRegionTrack",
+ rstarts = "exonStarts", rends = "exonEnds", gene = "name",</pre>
```

```
+ symbol = "name2", transcript = "name", strand = "strand",
+ fill = "#960000", name = "Ensembl Genes")
```

The CpG and SNP tracks are slightly different since a GeneRegionTrack representation would not be particularly useful. Instead, we can use AnnotationTrack objects as containers. The overall process using the UcscTrack meta-constructor remains the same.

```
> cpgIslands <- UcscTrack(genome = "mm9", chromosome = "chrX",
+ track = "cpgIslandExt", from = from, to = to,
+ trackType = "AnnotationTrack", start = "chromStart",
+ end = "chromEnd", id = "name", shape = "box",
+ fill = "#006400", name = "CpG Islands")
> snpLocations <- UcscTrack(genome = "mm9", chromosome = "chrX",
+ track = "snp128", from = from, to = to, trackType = "AnnotationTrack",
+ start = "chromStart", end = "chromEnd", id = "name",
+ feature = "func", strand = "strand", shape = "box",
+ stacking = "dense", fill = "black", name = "SNPs")</pre>
```

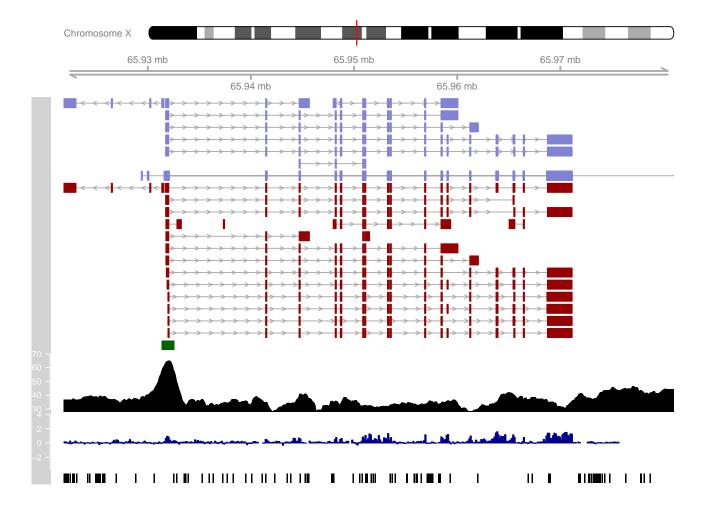
Most of UCSC's *DataTrack*-like tracks are a little more complex and represent a collection of several subtracks, with data originating from multiple tables. To make sure that we get the correct information we have to be a little bit more specific here and also define the particular table on the UCSC data base to use.

```
> conservation <- UcscTrack(genome = "mm9", chromosome = "chrX",
+ track = "Conservation", table = "phyloP30wayPlacental",
+ from = from, to = to, trackType = "DataTrack",
+ start = "start", end = "end", data = "score",
+ type = "hist", window = "auto", col.histogram = "darkblue",
+ fill.histogram = "darkblue", ylim = c(-3.7, 4),
+ name = "Conservation")
> gcContent <- UcscTrack(genome = "mm9", chromosome = "chrX",
+ track = "GC Percent", table = "gc5Base", from = from,
+ to = to, trackType = "DataTrack", start = "start",
+ end = "end", data = "score", type = "hist", window = -1,
+ windowSize = 1500, fill.histogram = "black",
+ col.histogram = "black", ylim = c(30, 70), name = "GC Percent")</pre>
```

To add some reference points we also need a genome axis and an *IdeogramTrack* of the x chromosome.

```
> axTrack <- GenomeAxisTrack()
> idxTrack <- IdeogramTrack(genome = "mm9", chromosome = "chrX")
   And finally we can plot all of our tracks.
> plotTracks(list(idxTrack, axTrack, knownGenes, refGenes, ensGenes, cpgIslands, gcContent, conservation,
```

snpLocations), from = from, to = to, showTitle = FALSE)



# 5 Bioconductor integration and file support

This short section is supposed to give a very brief overview over the different track classes in the Gviz package and how those can be constructed from the typical Bioconductor classes that deal with genomic data. The list ist by no means complete, and a closer look at a track class' documentation should provide all the possible options.

Gviz class	Bioconductor class	Method		
AnnotationTrack	data.frame	Constructor		
	IRanges	Constructor + additional arguments		
	GRanges	Constructor or setAs method, additional data in element-		
		Metadata		
	GRangesList	Constructor or setAs method		
GeneRegionTrack	data.frame	Constructor		
	IRanges	Constructor + additional arguments		
	GRanges	Constructor or setAs method, additional data in element-		
		Metadata		
	GRangesList	Constructor or setAs method, additional data in element-		
		Metadata		
	TranscriptDb	Constructor or setAs method		

DataTrack	data.frame	Constructor		
	IRanges	Constructor + additional data matrix		
	GRanges	Constructor or setAs method, numeric data in element-		
		Metadata		
SequenceTrack	DNAStringSet	Constructor		
	BSgenome	Constructor		

We have seen in previous examples that the Gviz package supports quite a wide range of standard file types as inputs for the class' constructor functions. In this section we summarize the file support for the different track classes in one place.

Gviz class	File type	Extension	Streaming	Details
AnnotationTrack	BED	.bed	_	Genomic locations from the manda-
				tory chrom, chromStart and chromEnd
				fields, and optionally the strand from
				the strand field. If present, the infor-
				mation in the name field is mapped to
				track item ids, and itemRgb is mapped
				to track item feature type. All other
	G			fields are currently ignored.
	GFF	.gff, .gff1	_	Only the following basic GFF fields are
				recognized: seqname, start, end, end,
				strand, feature (mapped to track
				item feature type) and group (to allow
	CEES			for track item grouping).
	GFF2	.gff2	_	Same as above, but feature grouping
				information may be provided either as
				Group or Parent attribute. Feature ids
				are mapped to one of the ID, Name or
	GFF3	mff2		Alias attributes.
	Grra	.gff3	_	Same as above, but feature grouping information has to be provided as the
				Parent attribute.
	BAM	.bam	. /	Only start and end locations as well
	DAM	. Dain	$\sqrt{}$	as the strand information for the reads
				are used. Read identifiers are used for
				track item grouping.
GeneRegionTrack	GTF	.gtf	_	A somewhat looser format definition
G 011010001011110011		1802		for gtf files is applied here where gene,
				transcript and exon identifiers and
				names can be parsed from the gene_id,
				gene_name, transcript_id, tran-
				script_name, exon_id or exon_id at-
				tributes.
	GFF	.gff, .gff1	_	This only supports very limited item
				grouping and thus complete gene mod-
				els can not be properly encoded.
	I	I	1	* * *

	GFF2	.gff2	_	In most instances this is identical to the GTF standard and it could make sense
				to rename the file accordingly.
	GFF3	.gff3	_	The gene-to-transcript and transcript-
				to-exon relationships are encoded in
				the parent and type attributes and the
				parser tries to accommodate most of the
				exisiting GFF3 variants.
DataTrack	BedGraph	.bedGraph	_	
	WIG	.wig	_	
	BigWig	.bigWig, .bw		
	BAM	.bam		Read coverage only is extracted from
				the bam file.
SequenceTrack	FASTA	.fa, fasta	(√)	Streaming only possible if an index file
				is found in the same directory as the
				original fasta file.
	2Bit	.2bit		

# 6 Composite plots for multiple chromosomes

As mentioned in the introduction section, a set of Gviz tracks has to share the same chromosome when plotted, i.e., only a single chromosome can be active during a given plotting operation. Consequently, we can not directly create plots for multiple chromosomes in a single call to the plotTracks function. However, since the underlying graphical infrastructure of the Gviz package uses grid graphics, we can build our own composite plot using multiple consecutive plotTracks calls. All we need to take care of is an adequate layout structure to plot into, and we also need to tell plotTracks not to clear the graphics device before plotting, which can be archieved by setting the function's add argument to FALSE. For details on how to create a layout structure in the grid graphics system, please see the help page at ?grid.layout).

We start by creating an *AnnotationTrack* objects and a *DataTrack* object which both contain data for several chromosomes.

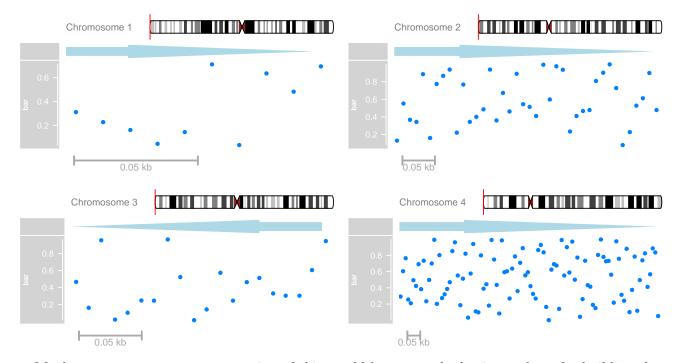
```
> chroms <- c("chr1", "chr2", "chr3", "chr4")
> maTrack <- AnnotationTrack(range = GRanges(seqnames = chroms,
+ ranges = IRanges(start = 1, width = c(100, 400,
+ 200, 1000)), strand = c("+", "+", "-", "+")),
+ genome = "mm9", chromosome = "chr1", name = "foo")
> mdTrack <- DataTrack(range = GRanges(seqnames = rep(chroms,
+ c(10, 40, 20, 100)), ranges = IRanges(start = c(seq(1,
+ 100, len = 10), seq(1, 400, len = 40), seq(1,
+ 200, len = 20), seq(1, 1000, len = 100)), width = 9),
+ values = runif(170)), data = "values", chromosome = "chr1",
+ genome = "mm9", name = "bar")</pre>
```

Now we also want a genome axis and an *IdeogramTrack* object to indicate the genomic context.

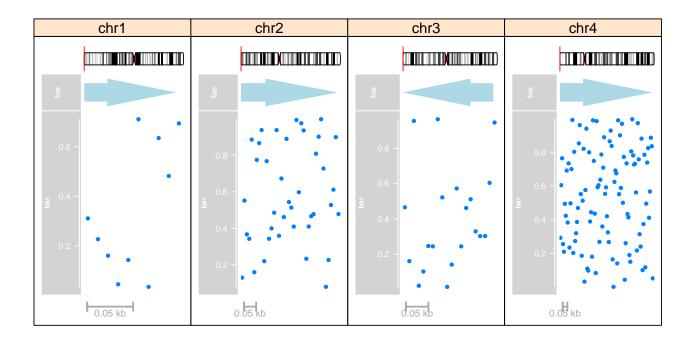
```
> mgTrack <- GenomeAxisTrack(scale = 50, labelPos = "below",
+ exponent = 3)
> chromosome(itrack) <- "chr1"</pre>
```

Finaly, we build a layout in which the plots for each chromosome are placed in a rectangular grid and repeatedly call plotTracks for each chromosome.

```
> ncols <- 2
> nrows <- length(chroms)%/%ncols
> grid.newpage()
> pushViewport(viewport(layout = grid.layout(nrows,
+ ncols)))
> for (i in seq_along(chroms)) {
+ pushViewport(viewport(layout.pos.col = ((i -
+ 1)%/ncols) + 1, layout.pos.row = (((i) -
+ 1)%//ncols) + 1))
+ plotTracks(list(itrack, maTrack, mdTrack, mgTrack),
+ chromosome = chroms[i], add = TRUE)
+ popViewport(1)
+ }
```



Maybe an even more compact version of this would be to use the lattice package for building the actual trellis, with plotTracks as the panel function.



## SessionInfo

[13] dichromat\_2.0-0

[16] munsell\_0.4.2

[19] scales\_0.2.3

[22] tools\_3.0.2

The following is the session info that generated this vignette:

```
> sessionInfo()
R version 3.0.2 (2013-09-25)
Platform: x86_64-unknown-linux-gnu (64-bit)
locale:
 [1] LC_CTYPE=en_US.UTF-8
                                 LC_NUMERIC=C
 [3] LC_TIME=en_US.UTF-8
                                 LC_COLLATE=C
 [5] LC_MONETARY=en_US.UTF-8
                                 LC_MESSAGES=en_US.UTF-8
 [7] LC_PAPER=en_US.UTF-8
                                 LC_NAME=C
 [9] LC_ADDRESS=C
                                 LC_TELEPHONE=C
[11] LC_MEASUREMENT=en_US.UTF-8 LC_IDENTIFICATION=C
attached base packages:
[1] parallel grid
                                   graphics grDevices utils
                        stats
[7] datasets methods
                         base
other attached packages:
 [1] lattice_0.20-24
 [2] GenomicFeatures_1.14.0
 [3] AnnotationDbi_1.24.0
 [4] Biobase_2.22.0
 [5] rtracklayer_1.22.0
 [6] BSgenome.Hsapiens.UCSC.hg19_1.3.19
 [7] BSgenome_1.30.0
 [8] Biostrings_2.30.0
 [9] GenomicRanges_1.14.0
[10] XVector_0.2.0
[11] IRanges_1.20.0
[12] BiocGenerics_0.8.0
[13] Gviz_1.6.0
[14] xtable_1.7-1
loaded via a namespace (and not attached):
                         Hmisc_3.12-2
 [1] DBI_0.2-7
                                              RColorBrewer_1.0-5
 [4] RCurl_1.95-4.1
                          RSQLite_0.11.4
                                              Rsamtools_1.14.0
 [7] XML_3.98-1.1
                          biomaRt_2.18.0
                                              biovizBase_1.10.0
[10] bitops_1.0-6
                          cluster_1.14.4
                                              colorspace_1.2-4
```

labeling\_0.2

stats4\_3.0.2

zlibbioc\_1.8.0

plyr\_1.8

latticeExtra\_0.6-26

rpart\_4.1-3

stringr\_0.6.2