Graphics and Data Visualization in R

Data Analysis in Genome Biology GEN242

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Overview

Graphics Environments

Base Graphics Grid Graphics lattice ggplot2

Specialty Graphics

Genome Graphics

ggbio

Additional Genome Graphics

Genome Browser: IGV

Outline

Overview

Graphics Environments
Base Graphics
Grid Graphics
lattice
ggplot2

Specialty Graphics

Genome Graphics ggbio Additional Genome Graphic

Genome Browser: IGV

Graphics in R

- Powerful environment for visualizing scientific data
- Integrated graphics and statistics infrastructure
- Publication quality graphics
- Fully programmable
- Highly reproducible
- Full LATEX Link & Sweave Link support
- Vast number of R packages with graphics utilities

Documentation on Graphics in R

General

- Graphics Task Page Link
- R Graph Gallery Link
- R Graphical Manual Link
- Paul Murrell's book R (Grid) Graphics Link

Interactive graphics

- rggobi (GGobi) Link
- iplots Link
- Open GL (rgl) Link

Graphics Environments

Viewing and saving graphics in R

- On-screen graphics
- postscript, pdf, svg
- jpeg/png/wmf/tiff/...

Four major graphic environments

- Low-level infrastructure
 - R Base Graphics (low- and high-level)
 - grid: Manual Link, Book Link
- High-level infrastructure
 - lattice: Manual Link, Intro Link, Book Link
 - ggplot2: Manual Link, Intro Link, Book Link

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Base Graphics: Overview

Important high-level plotting functions

- plot: generic x-y plotting
- barplot: bar plots
- boxplot: box-and-whisker plot
- hist: histograms
- pie: pie charts
- dotchart: cleveland dot plots
- image, heatmap, contour, persp: functions to generate image-like plots
- qqnorm, qqline, qqplot: distribution comparison plots
- pairs, coplot: display of multivariant data

Help on these functions

- ?myfct
- ?plot
- ?par

Base Graphics: Preferred Input Data Objects

- Matrices and data frames
- Vectors
- Named vectors

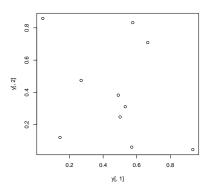
Scatter Plot: very basic

Sample data set for subsequent plots

```
> set.seed(1410)
```

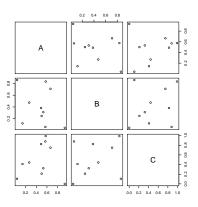
 $> y <- \ {\tt matrix(runif(30), ncol=3, dimnames=list(letters[1:10], LETTERS[1:3]))} \\$

> plot(y[,1], y[,2])



Scatter Plot: all pairs

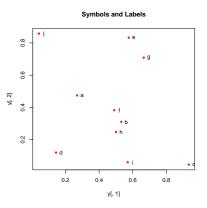
> pairs(y)



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Scatter Plot: with labels

> plot(y[,1], y[,2], pch=20, col="red", main="Symbols and Labels") > text(y[,1]+0.03, y[,2], rownames(y))



Scatter Plots: more examples

Print instead of symbols the row names

Important arguments

- mar: specifies the margin sizes around the plotting area in order: c(bottom, left, top, right)
- col: color of symbols
- pch: type of symbols, samples: example(points)
- 1wd: size of symbols
- cex.*: control font sizes
- For details see ?par

Scatter Plots: more examples

```
Add a regression line to a plot

> plot(y[,1], y[,2])

> myline <- lm(y[,2]~y[,1]); abline(myline, lwd=2)

> summary(myline)

Same plot as above, but on log scale

> plot(y[,1], y[,2], log="xy")

Add a mathematical expression to a plot

> plot(y[,1], y[,2]); text(y[1,1], y[1,2],

> expression(sum(frac(1,sqrt(x^2*pi)))), cex=1.3)
```

Exercise 1 Scatter Plots

- Task 1 Generate scatter plot for first two columns in iris data frame and color dots by its Species column.
- Task 2 Use the xlim/ylim arguments to set limits on the x- and y-axes so that all data points are restricted to the left bottom quadrant of the plot.

Structure of iris data set:

- > class(iris)
- [1] "data.frame"
- > iris[1:4.]

Sepal.Length Sepal.Width Petal.Length Petal.Width Species

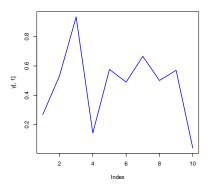
| 1 | 5.1 | 3.5 | 1.4 | 0.2 | setosa |
|---|-----|-----|-----|-----|--------|
| 2 | 4.9 | 3.0 | 1.4 | 0.2 | setosa |
| 3 | 4.7 | 3.2 | 1.3 | 0.2 | setosa |
| 4 | 4.6 | 3.1 | 1.5 | 0.2 | setosa |

> table(iris\$Species)

```
setosa versicolor virginica
   50
               50
                          50
```

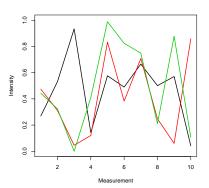
Line Plot: Single Data Set

> plot(y[,1], type="1", lwd=2, col="blue")



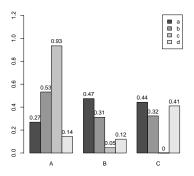
Graphics and Data Visualization in R

Line Plots: Many Data Sets



Bar Plot Basics

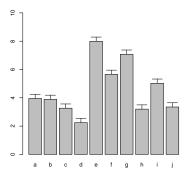
```
> barplot(y[1:4,], ylim=c(0, max(y[1:4,])+0.3), beside=TRUE,
+ legend=letters[1:4])
> text(labels=round(as.vector(as.matrix(y[1:4,])),2), x=seq(1.5, 13, by=1)
+ sort(rep(c(0,1,2), 4)), y=as.vector(as.matrix(y[1:4,]))+0.04)
```



Bar Plots with Error Bars

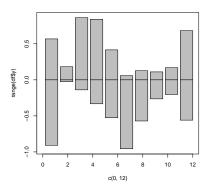
```
> bar <- barplot(m <- rowMeans(y) * 10, ylim=c(0, 10)) > stdev <- sd(t(y))
```

> arrows(bar, m, bar, m + stdev, length=0.15, angle = 90)



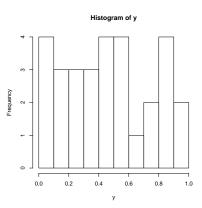
Mirrored Bar Plots

```
> df <- data.frame(group = rep(c("Above", "Below"), each=10), x = rep(1:10, 2),
> plot(c(0,12),range(df$y),type = "n")
> barplot(height = df$y[df$group == "Above"], add = TRUE,axes = FALSE)
> barplot(height = df$v[df$group == "Below"], add = TRUE,axes = FALSE)
```



Histograms

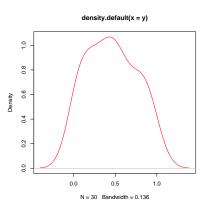
> hist(y, freq=TRUE, breaks=10)



Graphics and Data Visualization in R Graphics Environments

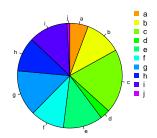
Density Plots

> plot(density(y), col="red")



Pie Charts

- > pie(y[,1], col=rainbow(length(y[,1]), start=0.1, end=0.8), clockwise=TRUE)
- > legend("topright", legend=row.names(y), cex=1.3, bty="n", pch=15, pt.cex=1.8,
- + col=rainbow(length(y[,1]), start=0.1, end=0.8), ncol=1)



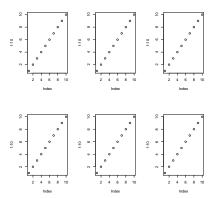
Color Selection Utilities

```
Default color palette and how to change it
> palette()
[1] "black" "red"
                         "green3" "blue"
                                              "cyan"
                                                       "magenta" "yellow" "gray
> palette(rainbow(5, start=0.1, end=0.2))
> palette()
[1] "#FF9900" "#FFBF00" "#FFE600" "#F2FF00" "#CCFF00"
> palette("default")
The gray function allows to select any type of gray shades by providing values from 0
to 1
> gray(seq(0.1, 1, by= 0.2))
[1] "#1A1A1A" "#4D4D4D" "#808080" "#B3B3B3" "#E6E6E6"
Color gradients with colorpanel function from gplots library
> library(gplots)
> colorpanel(5, "darkblue", "vellow", "white")
Much more on colors in R see Earl Glynn's color chart Link
```

Arranging Several Plots on Single Page

With par(mfrow=c(nrow,ncol)) one can define how several plots are arranged next to each other.

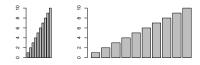
> par(mfrow=c(2,3)); for(i in 1:6) { plot(1:10) }

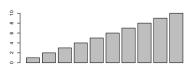


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Arranging Plots with Variable Width

The layout function allows to divide the plotting device into variable numbers of rows and columns with the column-widths and the row-heights specified in the respective arguments.





Saving Graphics to Files

After the pdf() command all graphs are redirected to file test.pdf. Works for all common formats similarly: jpeg, png, ps, tiff, ...

```
> pdf("test.pdf"); plot(1:10, 1:10); dev.off()
```

Generates Scalable Vector Graphics (SVG) files that can be edited in vector graphics programs, such as InkScape.

```
> svg("test.svg"); plot(1:10, 1:10); dev.off()
```

Exercise 2: Bar Plots

- Task 1 Calculate the mean values for the Species components of the first four columns in the iris data set. Organize the results in a matrix where the row names are the unique values from the iris Species column and the column names are the same as in the first four iris columns.
- Task 2 Generate two bar plots: one with stacked bars and one with horizontally arranged bars.

Structure of iris data set:

```
> class(iris)
```

[1] "data.frame"

> iris[1:4,]

```
Sepal.Length Sepal.Width Petal.Length Petal.Width Species
           5.1
                       3.5
                                    1.4
                                                0.2
                                                     setosa
2
           4.9
                       3.0
                                    1.4
                                                0.2 setosa
           4.7
                       3.2
                                    1.3
                                                0.2
                                                     setosa
           4.6
                       3.1
                                    1.5
                                                0.2
                                                     setosa
```

> table(iris\$Species)

setosa versicolor virginica 50 50 50

Outline

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Graphics Environments

Base Graphics

Grid Graphics

lattice ggplot2

Specialty Graphics

Genome Graphics ggbio Additional Genome Graphics

Genome Browser: IGV

grid Graphics Environment

- What is grid?
 - Low-level graphics system
 - Highly flexible and controllable system
 - Does not provide high-level functions
 - Intended as development environment for custom plotting functions
 - Pre-installed on new R distributions
- Documentation and Help
 - Manual Link
 - Book Link

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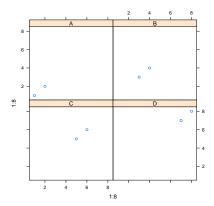
lattice Environment

- What is lattice?
 - High-level graphics system
 - Developed by Deepayan Sarkar
 - Implements Trellis graphics system from S-Plus
 - Simplifies high-level plotting tasks: arranging complex graphical features
 - Syntax similar to R's base graphics
- Documentation and Help
 - Manual Link
 - Intro Link
 - Book Link

 - library(help=lattice) opens a list of all functions available in the lattice package
 - Accessing and changing global parameters: ?lattice.options and ?trellis.device

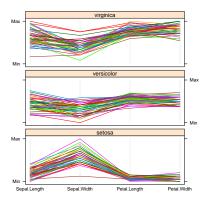
Scatter Plot Sample

```
> library(lattice)
> p1 <- xyplot(1:8 ~ 1:8 | rep(LETTERS[1:4], each=2), as.table=TRUE)
> plot(p1)
```



Line Plot Sample

```
> library(lattice)
> p2 <- parallelplot(~iris[1:4] | Species, iris, horizontal.axis = FALSE,
+ layout = c(1, 3, 1))
> plot(p2)
```



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ggplot2 Environment

- What is ggplot2?
 - High-level graphics system
 - Implements grammar of graphics from Leland Wilkinson Link
 - Streamlines many graphics workflows for complex plots
 - Syntax centered around main ggplot function
 - Simpler qplot function provides many shortcuts
- Documentation and Help
 - Manual Link
 - Intro Link
 - Book (Link)

 - Cookbook for R Link

ggplot2 Usage

- ggplot function accepts two arguments
 - Data set to be plotted
 - Aesthetic mappings provided by aes function
- Additional parameters such as geometric objects (e.g. points, lines, bars) are passed on by appending them with + as separator.
- List of available geom_* functions: Link
- Settings of plotting theme can be accessed with the command theme_get() and its settings can be changed with theme().
- Preferred input data object
 - ullet qplot: data.frame (support for vector, matrix, ...)
 - ggplot: data.frame
- Packages with convenience utilities to create expected inputs
 - plyr
 - reshape

qplot Function

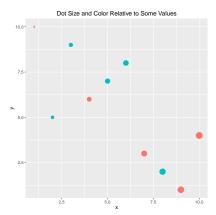
- qplot syntax is similar to R's basic plot function
- Arguments:
 - x: x-coordinates (e.g. col1)
 - y: y-coordinates (e.g. col2)
 - data: data frame with corresponding column names
 - xlim, ylim: *e.g.* xlim=c(0,10)
 - log: e.g. log="x" or log="xy"
 - main: main title; see ?plotmath for mathematical formula
 - xlab, ylab: labels for the x- and y-axes
 - color, shape, size
 - ...: many arguments accepted by plot function

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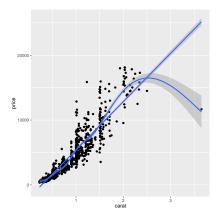
qplot: Scatter Plots

```
Create sample data
> library(ggplot2)
> x < - sample(1:10, 10); y < - sample(1:10, 10); cat < - rep(c("A", "B"), 5)
Simple scatter plot
> qplot(x, y, geom="point")
Prints dots with different sizes and colors
> qplot(x, y, geom="point", size=x, color=cat,
        main="Dot Size and Color Relative to Some Values")
Drops legend
> qplot(x, y, geom="point", size=x, color=cat) +
        theme(legend.position = "none")
Plot different shapes
> qplot(x, y, geom="point", size=5, shape=cat)
```

qplot: Scatter Plot with qplot

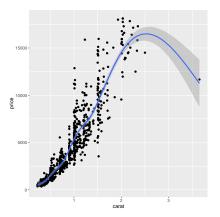


qplot: Scatter Plot with Regression Line



qplot: Scatter Plot with Local Regression Curve (loess)

```
> p <- qplot(carat, price, data=dsmall, geom=c("point", "smooth"))
> print(p) # Setting se=FALSE removes error shade
```



ggplot Function

- More important than qplot to access full functionality of ggplot2
- Main arguments
 - data set, usually a data.frame
 - aesthetic mappings provided by aes function
- General ggplot syntax

```
• ggplot(data, aes(...)) + geom_*() + ... + stat_*() + ...
```

- Layer specifications
 - geom_*(mapping, data, ..., geom, position)
 - stat_*(mapping, data, ..., stat, position)
- Additional components
 - scales
 - coordinates
 - facet.
- aes() mappings can be passed on to all components (ggplot, geom_*, etc.).
 Effects are global when passed on to ggplot() and local for other components.
 - x. v
 - color: grouping vector (factor)
 - group: grouping vector (factor)

Changing Plotting Themes with ggplot

- Theme settings can be accessed with theme_get()
- Their settings can be changed with theme()
- Some examples
 - Change background color to white
 - ... + theme(panel.background=element_rect(fill = "white", colour = "black"))

Graphics and Data Visualization in R Graphics Environments ggplot2 Slide 45/94

Storing ggplot Specifications

```
Plots and layers can be stored in variables
> p <- ggplot(dsmall, aes(carat, price)) + geom_point()
> p # or print(p)
Returns information about data and aesthetic mappings followed by each layer
> summary(p)
Prints dots with different sizes and colors
> bestfit <- geom_smooth(methodw = "lm", se = F, color = alpha("steelblue", 0.5)</pre>
```

Syntax to pass on other data sets

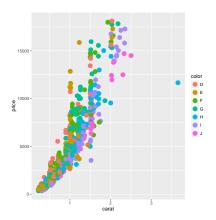
```
> p %+% diamonds[sample(nrow(diamonds), 100),]
```

> p + bestfit # Plot with custom regression line

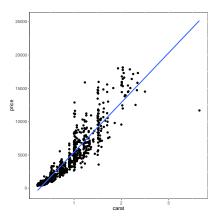
Saves plot stored in variable p to file

```
> ggsave(p, file="myplot.pdf")
```

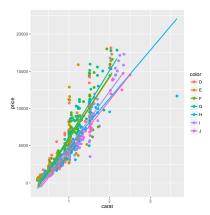
ggplot: Scatter Plot



ggplot: Scatter Plot with Regression Line

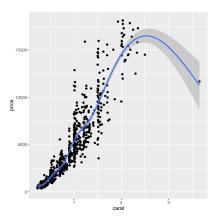


ggplot: Scatter Plot with Several Regression Lines



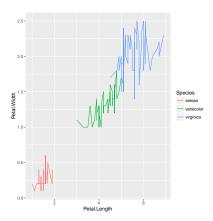
ggplot: Scatter Plot with Local Regression Curve (loess)

```
> p <- ggplot(dsmall, aes(carat, price)) + geom_point() + geom_smooth()
> print(p) # Setting se=FALSE removes error shade
```

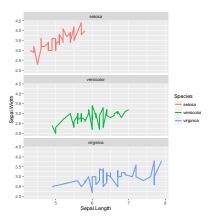


ggplot: Line Plot

```
> p <- ggplot(iris, aes(Petal.Length, Petal.Width, group=Species,
+ color=Species)) + geom_line()
> print(p)
```



ggplot: Faceting



Exercise 3: Scatter Plots

- Task 1 Generate scatter plot for first two columns in iris data frame and color dots by its Species column.
- Task 2 Use the xlim, ylim functionss to set limits on the x- and y-axes so that all data points are restricted to the left bottom quadrant of the plot.
- Task 3 Generate corresponding line plot with faceting show individual data sets in saparate plots.

Structure of iris data set:

- > class(iris)
- [1] "data.frame"
- > iris[1:4,]

Sepal.Length Sepal.Width Petal.Length Petal.Width Species

| | bepar. Length | bcpar.wrach | 1 Cuar. Length | I Coar. wrach | pheeres |
|---|---------------|-------------|----------------|---------------|---------|
| 1 | 5.1 | 3.5 | 1.4 | 0.2 | setosa |
| 2 | 4.9 | 3.0 | 1.4 | 0.2 | setosa |
| 3 | 4.7 | 3.2 | 1.3 | 0.2 | setosa |
| 4 | 4.6 | 3.1 | 1.5 | 0.2 | setosa |

> table(iris\$Species)

setosa versicolor virginica 50 50 50

ggplot: Bar Plots

```
> iris_sd <- aggregate(iris[,1:4], by=list(Species=iris$Species), FUN=sd)
Convert iris_mean with melt
> library(reshape2) # Defines melt function
> df_mean <- melt(iris_mean, id.vars=c("Species"), variable.name = "Samples", value.
Convert iris_sd with melt
> df_sd <- melt(iris_sd, id.vars=c("Species"), variable.name = "Samples", value.name
Define standard deviation limits</pre>
```

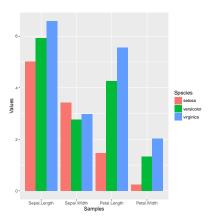
> limits <- aes(ymax = df_mean[, "Values"] + df_sd[, "Values"], ymin=df_mean[, "Values"

Sample Set: the following transforms the iris data set into a ggplot2-friendly format.

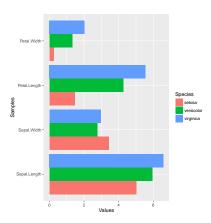
Calculate mean values for aggregates given by Species column in iris data set

> iris_mean <- aggregate(iris[,1:4], by=list(Species=iris\$Species), FUN=mean)
Calculate standard deviations for aggregates given by Species column in iris data set</pre>

ggplot: Bar Plot

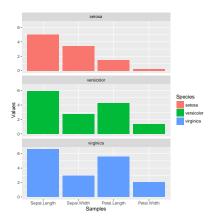


ggplot: Bar Plot Sideways

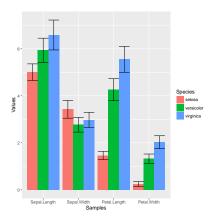


ggplot: Bar Plot with Faceting

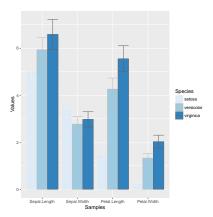
```
> p <- ggplot(df_mean, aes(Samples, Values)) + geom_bar(aes(fill = Species), sta
+ facet_wrap(~Species, ncol=1)
> print(p)
```



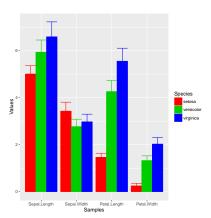
ggplot: Bar Plot with Error Bars



ggplot: Changing Color Settings

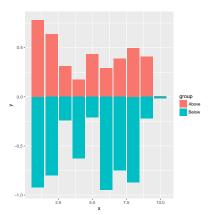


ggplot: Using Standard Colors



ggplot: Mirrored Bar Plots

```
> df <- data.frame(group = rep(c("Above", "Below"), each=10), x = rep(1:10, 2), y = c(runif(10, 0, 1), runit) > p <- ggplot(df, aes(x=x, y=y, fill=group)) + + geom_bar(stat="identity", position="identity") > print(p)
```



Exercise 4: Bar Plots

- Task 1 Calculate the mean values for the Species components of the first four columns in the iris data set. Use the melt function from the *reshape2* package to bring the results into the expected format for ggplot.
- Task 2 Generate two bar plots: one with stacked bars and one with horizontally arranged bars.

Structure of iris data set:

```
> class(iris)
```

[1] "data.frame"

> iris[1:4,]

Sepal.Length Sepal.Width Petal.Length Petal.Width Species

| 1 | 5 | .1 | ; | 3.5 | 1.4 | 0.2 | setosa |
|---|---|----|---|-----|-----|-----|--------|
| 2 | 4 | .9 | ; | 3.0 | 1.4 | 0.2 | setosa |
| 3 | 4 | .7 | ; | 3.2 | 1.3 | 0.2 | setosa |
| 4 | 4 | .6 | ; | 3.1 | 1.5 | 0.2 | setosa |

> table(iris\$Species)

setosa versicolor virginica 50 50 50

ggplot: Data Reformatting Example for Line Plot

```
> y <- matrix(rnorm(500), 100, 5, dimnames=list(paste("g", 1:100, sep=""), paste("Sample", 1:5, sep="")))
> y <- data.frame(Position=1:length(y[,1]), y)
> y[1:4, ] # First rows of input format expected by melt()
Position Sample1 Sample2 Sample3 Sample4 Sample5
g1 1 1.0002088 0.6850199 -0.21324932 1.27195056 1.0479301
g2 2 -1.2024596 -1.5004962 -0.01111579 0.07584497 -0.7100662
```

> df <- melt(y, id.vars=c("Position"), variable.name = "Samples", value.name="Values")
> p <- ggplot(df, aes(Position, Values)) + geom_line(aes(color=Samples)) + facet_wrap(~Samples, ncol=1)
> print(p)

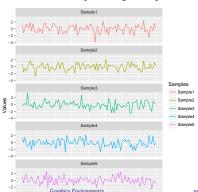
1.1084695

1.7139697

- > ## Represent same data in box plot
- > ## ggplot(df, aes(Samples, Values, fill=Samples)) + geom_boxplot()

0.1023678 -0.5153367 0.28564390 1.41522878

1.3294248 -1.2084007 -0.19581898 -0.42361768

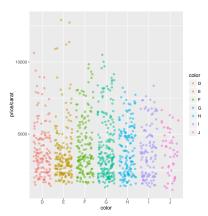


Graphics and Data Visualization in R

g3

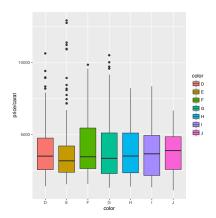
g4

ggplot: Jitter Plots



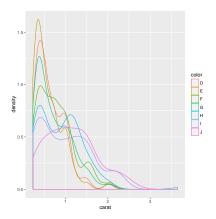
ggplot: Box Plots

> p <- ggplot(dsmall, aes(color, price/carat, fill=color)) + geom_boxplot()
> print(p)



ggplot: Density Plot with Line Coloring

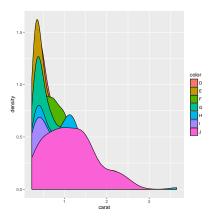
```
> p <- ggplot(dsmall, aes(carat)) + geom_density(aes(color = color))
> print(p)
```



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ggplot: Density Plot with Area Coloring

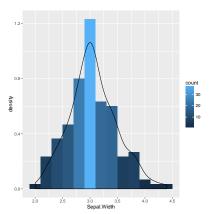
```
> p <- ggplot(dsmall, aes(carat)) + geom_density(aes(fill = color))
> print(p)
```



Graphics and Data Visualization in R Graphics Environments ggplot2 Slide 67/94

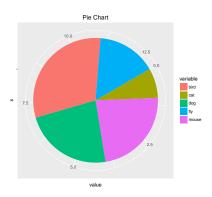
ggplot: Histograms

```
> p <- ggplot(iris, aes(x=Sepal.Width)) + geom_histogram(aes(y = ..density..,
+ fill = ..count..), binwidth=0.2) + geom_density()
> print(p)
```



ggplot: Pie Chart

```
> df <- data.frame(variable=rep(c("cat", "mouse", "dog", "bird", "fly")),
+ value=c(1,3,3,4,2))
> p <- ggplot(df, aes(x = "", y = value, fill = variable)) +
+ geom_bar(width = 1, stat="identity") +
+ coord_polar("y", start=pi / 3) + ggtitle("Pie Chart")
> print(p)
```



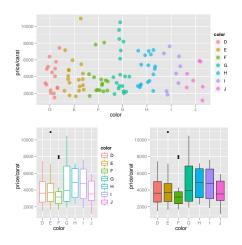
ggplot: Wind Rose Pie Chart



ggplot: Arranging Graphics on One Page

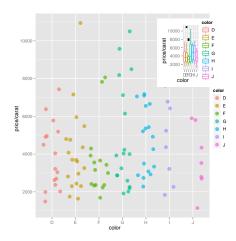
```
> library(grid)
> a <-ggplot(dsmall, aes(color, price/carat)) + geom_jitter(size=4, alpha = I(1 / 1.5), aes(color=color))
> b <-ggplot(dsmall, aes(color, price/carat, color=color)) + geom_boxplot()
> c <-ggplot(dsmall, aes(color, price/carat, fill=color)) + geom_boxplot() + theme(legend.position = "none > grid.newpage() # Open a new page on grid device
> pushViewport(viewport(layout = grid.layout(2, 2))) # Assign to device viewport with 2 by 2 grid layout
> print(a, vp = viewport(layout.pos.row = 1, layout.pos.col = 1:2))
> print(b, vp = viewport(layout.pos.row = 2, layout.pos.col = 2. width=0.3. height=0.3. x=0.8. v=0.8))
```

ggplot: Arranging Graphics on One Page



ggplot: Inserting Graphics into Plots

- > # pdf("insert.pdf")
- > print(a)
- > print(b, vp=viewport(width=0.3, height=0.3, x=0.8, y=0.8))
- > # dev.off()



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Specialty Graphics

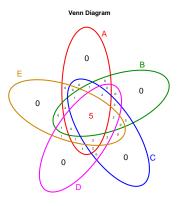
Genome Graphics ggbio Additional Genome Graphic

Venn Diagrams (Code)

> # dev.off()

```
> source("http://faculty.ucr.edu/~tgirke/Documents/R_BioCond/My_R_Scripts/overLapper.R")
> setlist5 <- list(A=sample(letters, 18), B=sample(letters, 16), C=sample(letters, 20), D=sample(letters, 20)
> CLlist5 <- overLapper(setlist=setlist5, sep="_", type="vennsets")
> counts <- sapply(CLlist5$Venn_List, length)
> # pdf("venn.pdf")
> vennPlot(counts=counts, ccol=c(rep(1,30),2), lcex=1.5, ccex=c(rep(1.5,5), rep(0.6,25),1.5))
```

Venn Diagram (Plot)



Unique objects: All = 26; S1 = 18; S2 = 16; S3 = 20; S4 = 22; S5 = 18

Figure: Venn Diagram

Compound Depictions with ChemmineR

- > library(ChemmineR)
- > data(sdfsample)
- > plot(sdfsample[1], print=FALSE)

CMP1

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ggbio: A Programmable Genome Browser

- A genome browser is a visualization tool for plotting different types of genomic data in separate tracks along chromosomes.
- The ggbio package (?) facilitates plotting of complex genome data objects, such as read alignments (SAM/BAM), genomic context/annotation information (gff/txdb), variant calls (VCF/BCF), and more. To easily compare these data sets, it extends the faceting facility of ggplot2 to genome browser-like tracks.
- Most of the core object types for handling genomic data with R/Bioconductor are supported: GRanges, GAlignments, VCF, etc. For more details, see Table 1.1 of the ggbio vignette Link.
- ggbio's convenience plotting function is autoplot. For more customizable plots, one can use the generic ggplot function.
- Apart from the standard ggplot2 plotting components, ggbio defines serval new components useful for genomic data visualization. A detailed list is given in Table 1.2 of the vignette Link.
- Useful web sites:
 - ggbio manual Link
 - ggbio functions Link
 - autoplot demo Link

Graphics and Data Visualization in R Genome Graphics ggbio Slide 80/94

Tracks: Aligning Plots Along Chromosomes

> df1 <- data.frame(time = 1:100, score = sin((1:100)/20)*10)
> p1 <- qplot(data = df1, x = time, y = score, geom = "line")</pre>

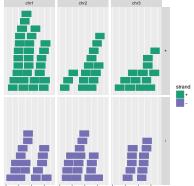
> library(ggbio)

```
> df2 <- data.frame(time = 30:120, score = sin((30:120)/20)*10, value = rnorm(120-30 +1))
> p2 <- ggplot(data = df2, aes(x = time, y = score)) + geom_line() + geom_point(size = 2, aes(color = value
> tracks(time1 = p1, time2 = p2) + xlim(1, 40) + theme_tracks_sunset()
                                10-
                               -10 -
```

Plotting Genomic Ranges

GRanges objects are essential for storing alignment or annotation ranges in R/Bioconductor. The following creates a sample GRanges object and plots its content.

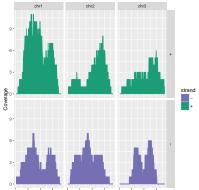
- > library(GenomicRanges)
- > set.seed(1); N <- 100; gr <- GRanges(seqnames = sample(c("chr1", "chr2", "chr3"), size = N, replace = TRU
- > autoplot(gr, aes(color = strand, fill = strand), facets = strand ~ seqnames)



0 bp 100 bp200 bp300 bp 0 bp 100 bp200 bp300 bp 0 bp 100 bp200 bp300 bp

Plotting Coverage Instead of Ranges

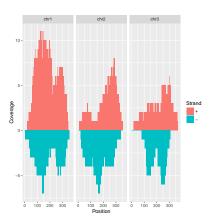
> autoplot(gr, aes(color = strand, fill = strand), facets = strand ~ seqnames, stat = "coverage")



0 bp 100 bp200 bp300 bp 0 bp 100 bp200 bp300 bp 0 bp 100 bp200 bp300 bp

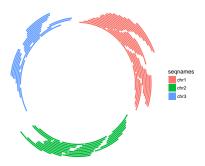
Graphics and Data Visualization in R Genome Graphics ggbio Slide 83/94

Mirrored Coverage Plot

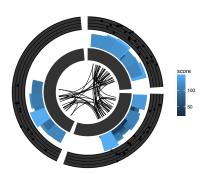


Circular Layout

> ggplot(gr) + layout_circle(aes(fill = seqnames), geom = "rect")



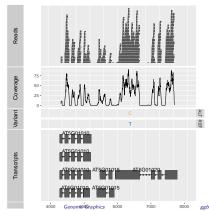
More Complex Circular Example



Viewing Alignments and Variants

To make the following example work, please download and unpack this data archive Link containing GFF, BAM and VCF sample files.

- > library(rtracklayer); library(GenomicFeatures); library(Rsamtools); library(GenomicAlignments); library(GenomicA
- > p1 <- autoplot(ga, geom = "rect")
- > p2 <- autoplot(ga, geom = "line", stat = "coverage")
- > vcf <- readVcf(file="data/varianttools_gnsap.vcf", genome="ATH1")
- > p3 <- autoplot(vcf[seqnames(vcf)=="Chr5"], type = "fixed") + xlim(4000, 8000) + theme(legend.position = "
- > txdb <- makeTxDbFromGFF(file="./data/TAIR10_GFF3_trunc.gff", format="gff3")
- > p4 <- autoplot(txdb, which=GRanges("Chr5", IRanges(4000, 8000)), names.expr = "gene_id")
- > tracks(Reads=p1, Coverage=p2, Variant=p3, Transcripts=p4, heights = c(0.3, 0.2, 0.1, 0.35)) + ylab("")



Additional Sample Plots

• autoplot demo Link

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Additional Genome Graphics

Additional Packages for Visualizing Genome Data

- Gviz Link
- RCircos (?) Link
- Genome Graphs Link
- genoPlotR Link

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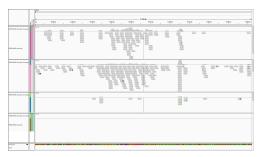
Alignment Viewing in IGV

View results in IGV

- Download and open IGV Link
- Select in menu in top left corner A. thaliana (TAIR10)
- Upload the following indexed/sorted Bam files with File -> Load from URL...

http://faculty.ucr.edu/~tgirke/HTML_Presentations/Manuals/Workshop_Dec_6_10_2012/Rrnaseq/results/SRR064154.fast http://faculty.ucr.edu/~tgirke/HTML_Presentations/Manuals/Workshop_Dec_6_10_2012/Rrnaseq/results/SRR064155.fast http://faculty.ucr.edu/~tgirke/HTML_Presentations/Manuals/Workshop_Dec_6_10_2012/Rrnaseq/results/SRR064166.fast http://faculty.ucr.edu/~tgirke/HTML_Presentations/Manuals/Workshop_Dec_6_10_2012/Rrnaseq/results/SRR064167.fast

To view area of interest, enter its coordinates Chr1:49,457-51,457 in position menu on top.



Create symbolic links for viewing BAM files in IGV

systemPipeR: utilities for building NGS analysis pipelines (Link)

```
library("systemPipeR")
symLink2bam(sysargs=args, htmldir=c(" /.html/", "somedir/"),
       urlbase="http://myserver.edu/ username/",
      urlfile="IGVurl.txt")
```

Bibliography I