

Programming in R

A Short Introduction

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Programming in R

Control Structures

Loops

Functions

Useful Utilities

Calling External Software

Running R Scripts

Building R Packages

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L^AT_EX

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Why Programming in R?

- Complete statistical package and programming language
- Reproducible research
- Efficient data structures make programming very easy
- Ease of implementing custom functions
- Powerful graphics
- Access to fast growing number of analysis packages
- Most widely used language in bioinformatics
- Is standard for data mining and biostatistical analysis
- Technical advantages: free, open-source, available for all OSs

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Examples

Important Operators

- Comparison operators
 - == (equal)
 - != (not equal)
 - > (greater than)
 - >= (greater than or equal)
 - < (less than)
 - <= (less than or equal)
- Logical operators
 - & (and)
 - | (or)
 - ! (not)

Conditional Executions: `if` statements

An `if` statement operates on length-one logical vectors.

Syntax

```
> if(TRUE) {  
+   statements_1  
+ } else {  
+   statements_2  
+ }
```

Example

```
> if(1==0) {  
+   print(1)  
+ } else {  
+   print(2)  
+ }
```

```
[1] 2
```

Conditional Executions: `ifelse` Statements

The `ifelse` statement operates on vectors.

Syntax

```
> ifelse(test, true_value, false_value)
```

Example

```
> x <- 1:10  
> ifelse(x<5, x, 0)  
  
[1] 1 2 3 4 0 0 0 0 0 0
```


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Examples

for Loops

Iterates over elements of a looping vector.

Syntax

```
> for(variable in sequence) {  
+     statements  
+ }
```

Example

```
> mydf <- iris  
> myve <- NULL  
> for(i in seq(along=mydf[,1])) {  
+     myve <- c(myve, mean(as.numeric(mydf[i,1:3])))  
+ }  
> myve[1:8]  
  
[1] 3.333333 3.100000 3.066667 3.066667 3.333333 3.666667 3.133333 3.300000
```

Inject into objects is much faster than append approach with `c`, `cbind`, etc.

```
> myve <- numeric(length(mydf[,1]))  
> for(i in seq(along=myve)) {  
+     myve[i] <- mean(as.numeric(mydf[i,1:3]))  
+ }  
> myve[1:8]  
  
[1] 3.333333 3.100000 3.066667 3.066667 3.333333 3.666667 3.133333 3.300000
```

Conditional Stop of Loops

The `stop` function can be used to break out of a loop (or a function) when a condition becomes TRUE and print an error message.

Example

```
> x <- 1:10
> z <- NULL
> for(i in seq(along=x)) {
+   if(x[i] < 5) {
+     z <- c(z, x[i]-1)
+   } else {
+     stop("values need to be <5")
+   }
+ }
```

while Loops

Iterates as long as a condition is true.

Syntax

```
> while(condition) {  
+     statements  
+ }
```

Example

```
> z <- 0  
> while(z<5) {  
+     z <- z + 2  
+     print(z)  
+ }
```

```
[1] 2
```

```
[1] 4
```

```
[1] 6
```

The apply Function Family: `apply`

Syntax

```
> apply(X, MARGIN, FUN, ARGS)
```

Arguments

- `X`: array, matrix or `data.frame`
- `MARGIN`: 1 for rows, 2 for columns
- `FUN`: one or more functions
- `ARGS`: possible arguments for functions

Example

```
> apply(iris[1:8,1:3], 1, mean)
```

```
      1      2      3      4      5      6      7      8  
3.333333 3.100000 3.066667 3.066667 3.333333 3.666667 3.133333 3.300000
```

The apply Function Family: `tapply`

Applies a function to vector components that are defined by a factor.

Syntax

```
> tapply(vector, factor, FUN)
```

Example

```
> iris[1:2,]
```

| | 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 |

```
> tapply(iris$Sepal.Length, iris$Species, mean)
```

| setosa | versicolor | virginica |
|--------|------------|-----------|
| 5.006 | 5.936 | 6.588 |

The apply Function Family: `sapply` and `lapply`

Both apply a function to vector or list objects. The function `lapply` always returns a list objects, while `sapply` tries to return vector or matrix objects when this is possible.

Examples

```
> x <- list(a = 1:10, beta = exp(-3:3), logic = c(TRUE,FALSE,FALSE,TRUE))
> lapply(x, mean)
```

```
$a
```

```
[1] 5.5
```

```
$beta
```

```
[1] 4.535125
```

```
$logic
```

```
[1] 0.5
```

```
> sapply(x, mean)
```

```
      a      beta      logic
5.500000 4.535125 0.500000
```

Often used in combination with a function definition

```
> lapply(names(x), function(x) mean(x))
```

```
> sapply(names(x), function(x) mean(x))
```

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Function Overview

A very useful feature of the R environment is the possibility to expand existing functions and to easily write custom functions. In fact, most of the R software can be viewed as a series of R functions.

Syntax to define functions

```
> myfct <- function(arg1, arg2, ...) {  
+   function_body  
+ }
```

Syntax to call functions

```
> myfct(arg1=..., arg2=...)
```

Function Syntax Rules

General Functions are defined by (1) assignment with the keyword `function`, (2) the declaration of arguments/variables (`arg1, arg2, ...`) and (3) the definition of operations (`function_body`) that perform computations on the provided arguments. A function name needs to be assigned to call the function.

Naming Function names can be almost anything. However, the usage of names of existing functions should be avoided.

Arguments It is often useful to provide default values for arguments (e.g.: `arg1=1:10`). This way they don't need to be provided in a function call. The argument list can also be left empty (`myfct <- function() fct_body`) when a function is expected to return always the same value(s). The argument `'...'` can be used to allow one function to pass on argument settings to another.

Body The actual expressions (commands/operations) are defined in the function body which should be enclosed by braces. The individual commands are separated by semicolons or new lines (preferred).

Usage Functions are called by their name followed by parentheses containing possible argument names. Empty parenthesis after the function name will result in an error message when a function requires certain arguments to be provided by the user. The function name alone will print the definition of a function.

Scope Variables created inside a function exist only for the life time of a function. Thus, they are not accessible outside of the function. To force variables in functions to exist globally, one can use the double assignment operator: `'<<-'`.

Function: Examples

Define sample function

```
> myfct <- function(x1, x2=5) {  
+     z1 <- x1 / x1  
+     z2 <- x2 * x2  
+     myvec <- c(z1, z2)  
+     return(myvec)  
+ }
```

Function usage

```
> ## Apply function to values 2 and 5  
> myfct(x1=2, x2=5)  
[1] 1 25  
  
> ## Run without argument names  
> myfct(2, 5)  
[1] 1 25  
  
> ## Makes use of default value '5'  
> myfct(x1=2)  
[1] 1 25  
  
> ## Print function definition  
> # myfct
```

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Debugging Utilities

Several debugging utilities are available for R. They include:

- `traceback`
- `browser`
- `options(error=recover)`
- `options(error=NULL)`
- `debug`

The Debugging in R page [Link](#) provides an overview of the available resources.

Regular Expressions

R's regular expression utilities work similar as in other languages. To learn how to use them in R, one can consult the main help page on this topic with `?regex`.

```
> ## The grep function can be used for finding patterns in strings, here letter  
> ## 'A' in vector 'month.name'.  
> month.name[grep("A", month.name)]
```

```
[1] "April" "August"
```

```
> ## Example for using regular expressions to substitute a pattern by another  
> ## one using a back reference. Remember: single escapes '\\' need to be double  
> ## escaped '\\\\' in R.  
> gsub('(i.*a)', 'xxx_\\1', "virginica", perl = TRUE)
```

```
[1] "vxxx_irginica"
```

Interpreting a Character String as Expression

Some useful examples

```
> ## Generates vector of object names in session  
> mylist <- ls()  
> ## Prints name of 1st entry  
> mylist[1]
```

```
[1] "i"
```

```
> ## Executes 1st entry as expression  
> get(mylist[1])
```

```
[1] 150
```

```
> # Alternative approach  
> eval(parse(text=mylist[1]))
```

```
[1] 150
```

Replacement, Split and Paste Functions for Strings

Selected examples

```
> ## Substitution with back reference which inserts in this example
> ## an '_' character
> x <- gsub("(a)", "\\1_", month.name[1], perl=T)
> x

[1] "Ja_nua_ry"

> ## Split string on inserted character from above
> strsplit(x, "_")

[[1]]
[1] "Ja" "nua" "ry"

> ## Reverse a character string by splitting first all characters
> ## into vector fields
> paste(rev(unlist(strsplit(x, NULL))), collapse="")

[1] "yr_aun_aJ"
```


Time, Date and Sleep

Selected examples

```
> ## Returns CPU (and other) times that an expression used (here ls)
> system.time(ls())
```

```
   user  system elapsed
    0      0         0
```

```
> ## Return the current system date and time
> date()
```

```
[1] "Thu Nov 29 16:48:34 2012"
```

```
> ## Pause execution of R expressions for a given number of
> ## seconds (e.g. in loop)
> Sys.sleep(1)
```

Import of Specific File Lines with Regular Expression

The following example demonstrates the retrieval of specific lines from an external file with a regular expression. First, an external file is created with the `cat` function, all lines of this file are imported into a vector with `readLines`, the specific elements (lines) are then retrieved with the `grep` function, and the resulting lines are split into vector fields with `strsplit`.

```
> cat(month.name, file="zzz.txt", sep="\n")
> x <- readLines("zzz.txt")
> x[1:6]

[1] "January" "February" "March"    "April"    "May"      "June"

> x <- x[c(grep("^J", as.character(x), perl = TRUE))]
> t(as.data.frame(strsplit(x, "u")))

      [,1] [,2]
c...Jan....ary.. "Jan" "ary"
c...J....ne..   "J"  "ne"
c...J....ly..   "J"  "ly"
```

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Run External Command-line Software

Example for running blastall from R

```
> system("blastall -p blastp -i seq.fasta -d uniprot -o seq.blastp")
```

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Options to Execute R Scripts

Executing R scripts from the R console

```
> source("my_script.R")
```

Execute an R script from command-line

```
Rscript my_script.R # or just ./myscript.R after making it executable
```

```
R CMD BATCH my_script.R # Alternative way 1
```

```
R --slave < my_script.R # Alternative way 2
```

Passing command-line arguments to R programs. In the given example the number 10 is passed on from the command-line as an argument to the R script which is used to return to STDOUT the first 10 rows of the iris sample data. If several arguments are provided, they will be interpreted as one string and need to be split in R with the `strsplit` function. A more detailed example can be found here: [Link](#)

```
## Create R script named 'test.R'  
#####  
myarg <- commandArgs()  
print(iris[1:myarg, ])  
#####  
  
## Then run it from the command-line  
Rscript test.R 10
```

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Short Overview of Package Building Process

Automatic package building with the `package.skeleton` function. The given example will create a directory named `mypackage` containing the skeleton of the package for all functions, methods and classes defined in the R script(s) passed on to the `code_files` argument. The basic structure of the package directory is described here: [Link](#). The package directory will also contain a file named `Read-and-delete-me` with instructions for completing the package:

```
> package.skeleton(name="mypackage", code_files=c("script1.R", "script2.R"),  
+                 namespace=TRUE)
```

Once a package skeleton is available one can build the package from the command-line (Linux/OS X). This will create a tarball of the package with its version number encoded in the file name. Subsequently, the package tarball needs to be checked for errors with:

```
R CMD build mypackage  
R CMD check mypackage_1.0.tar.gz
```

Install package from source

```
> install.packages("mypackage_1.0.tar.gz", repos=NULL)
```

For more details see here: [Link](#)

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Examples

Exercise 1: for loop

Task 1.1: Compute the mean of each row in `myMA` by applying the `mean` function in a `for` loop

```
> myMA <- matrix(rnorm(500), 100, 5, dimnames=list(1:100, paste("C", 1:5, sep="")))
> myve_for <- NULL
> for(i in seq(along=myMA[,1])) {
+   myve_for <- c(myve_for, mean(as.numeric(myMA[i, ])))
+ }
> myResult <- cbind(myMA, mean_for=myve_for)
> myResult[1:4, ]
```

| | C1 | C2 | C3 | C4 | C5 | mean_for |
|---|------------|------------|------------|------------|-------------|------------|
| 1 | -1.2075694 | -1.0365050 | -0.9283440 | 1.4959622 | -1.89454896 | -0.7142010 |
| 2 | 0.4437038 | 0.7256355 | 0.5402089 | -0.1669735 | 0.19199508 | 0.3469140 |
| 3 | -0.1423426 | 1.1938817 | 0.9793765 | -0.7811897 | -0.05497477 | 0.2389502 |
| 4 | 0.5841127 | -0.4491676 | 0.3313197 | 0.8148789 | -0.18115793 | 0.2199972 |

Exercise 1: while loop

Task 1.2: Compute the mean of each row in `myMA` by applying the mean function in a while loop

```
> z <- 1
> myve_while <- NULL
> while(z <= length(myMA[,1])) {
+   myve_while <- c(myve_while, mean(as.numeric(myMA[z, ])))
+   z <- z + 1
+ }
> myResult <- cbind(myMA, mean_for=myve_for, mean_while=myve_while)
> myResult[1:4, -c(1,2)]
```

| | C3 | C4 | C5 | mean_for | mean_while |
|---|------------|------------|-------------|------------|------------|
| 1 | -0.9283440 | 1.4959622 | -1.89454896 | -0.7142010 | -0.7142010 |
| 2 | 0.5402089 | -0.1669735 | 0.19199508 | 0.3469140 | 0.3469140 |
| 3 | 0.9793765 | -0.7811897 | -0.05497477 | 0.2389502 | 0.2389502 |
| 4 | 0.3313197 | 0.8148789 | -0.18115793 | 0.2199972 | 0.2199972 |

Task 1.3: Confirm that the results from both mean calculations are identical

```
> all(myResult[,6] == myResult[,7])
[1] TRUE
```

Exercise 1: apply loop and avoiding loops in R

Task 1.4: Compute the mean of each row in `myMA` by applying the `mean` function in an `apply` loop

```
> myve_apply <- apply(myMA, 1, mean)
> myResult <- cbind(myMA, mean_for=myve_for, mean_while=myve_while, mean_apply=myve_apply)
> myResult[1:4, -c(1,2)]
```

| | C3 | C4 | C5 | mean_for | mean_while | mean_apply |
|---|------------|------------|-------------|------------|------------|------------|
| 1 | -0.9283440 | 1.4959622 | -1.89454896 | -0.7142010 | -0.7142010 | -0.7142010 |
| 2 | 0.5402089 | -0.1669735 | 0.19199508 | 0.3469140 | 0.3469140 | 0.3469140 |
| 3 | 0.9793765 | -0.7811897 | -0.05497477 | 0.2389502 | 0.2389502 | 0.2389502 |
| 4 | 0.3313197 | 0.8148789 | -0.18115793 | 0.2199972 | 0.2199972 | 0.2199972 |

Task 1.5: When operating on large data sets it is much faster to use the `rowMeans` function

```
> mymean <- rowMeans(myMA)
> myResult <- cbind(myMA, mean_for=myve_for, mean_while=myve_while, mean_apply=myve_apply, mean_int=mymean)
> myResult[1:4, -c(1,2,3)]
```

| | C4 | C5 | mean_for | mean_while | mean_apply | mean_int |
|---|------------|-------------|------------|------------|------------|------------|
| 1 | 1.4959622 | -1.89454896 | -0.7142010 | -0.7142010 | -0.7142010 | -0.7142010 |
| 2 | -0.1669735 | 0.19199508 | 0.3469140 | 0.3469140 | 0.3469140 | 0.3469140 |
| 3 | -0.7811897 | -0.05497477 | 0.2389502 | 0.2389502 | 0.2389502 | 0.2389502 |
| 4 | 0.8148789 | -0.18115793 | 0.2199972 | 0.2199972 | 0.2199972 | 0.2199972 |

Exercise 2: functions

Task 2.1: Use the following code as basis to implement a function that allows the user to compute the mean for any combination of columns in a matrix or data frame. The first argument of this function should specify the input data set, the second the mathematical function to be passed on (e.g. mean, sd, max) and the third one should allow the selection of the columns by providing a grouping vector.

```
> myMA <- matrix(rnorm(100000), 10000, 10, dimnames=list(1:10000, paste("C", 1:10)))
> myMA[1:2,]

      C1      C2      C3      C4      C5      C6      C7
1 0.01918604 0.6259764 0.1408057 2.1601373 0.5487102 -0.8725235 -0.2845925 -1.0000000
2 0.23373842 0.5535612 0.5977537 -0.5298841 0.5000829 1.1534965 1.9919369 -1.0000000

> myList <- tapply(colnames(myMA), c(1,1,1,2,2,2,3,3,4,4), list)
> names(myList) <- sapply(myList, paste, collapse="_")
> myMAmean <- sapply(myList, function(x) apply(myMA[,x], 1, mean))
> myMAmean[1:4,]

      C1_C2_C3      C4_C5_C6      C7_C8      C9_C10
1 0.2619894 0.6121080 -0.7690928 0.3734742
2 0.4616844 0.3745651 0.4625598 -0.3640949
3 0.8763321 -0.4931097 -0.1746146 -0.6461481
4 0.1054418 -0.1076098 -0.3108006 -0.9934832
```

Exercise 3: nested loops to generate similarity matrices

Task 3.1: Create a sample list populated with character vectors of different lengths

```
> setlist <- lapply(11:30, function(x) sample(letters, x, replace=TRUE))  
> names(setlist) <- paste("S", seq(along=setlist), sep="")  
> setlist[1:6]
```

\$S1

```
[1] "p" "q" "g" "y" "x" "j" "d" "a" "b" "m" "q"
```

\$S2

```
[1] "i" "w" "g" "a" "z" "l" "j" "q" "t" "h" "e" "q"
```

\$S3

```
[1] "g" "l" "h" "f" "n" "e" "h" "j" "j" "l" "h" "o" "o"
```

\$S4

```
[1] "w" "r" "w" "t" "s" "w" "c" "r" "b" "i" "l" "q" "v" "o"
```

\$S5

```
[1] "f" "g" "t" "j" "c" "q" "t" "b" "d" "b" "h" "h" "v" "s" "v"
```

\$S6

```
[1] "t" "r" "z" "m" "v" "e" "r" "u" "d" "j" "g" "t" "x" "e" "q" "z"
```

Exercise 3: nested loops to generate similarity matrices

Task 3.2: Compute the length for all pairwise intersects of the vectors stored in `setlist`. The intersects can be determined with the `%in%` function like this:
`sum(setlist[[1]] %in% setlist[[2]])`

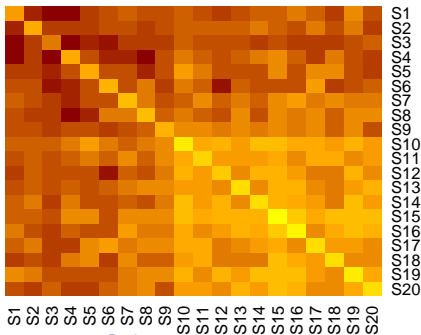
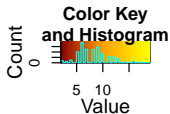
```
> setlist <- sapply(setlist, unique)
> olMA <- sapply(names(setlist), function(x) sapply(names(setlist),
+       function(y) sum(setlist[[x]] %in% setlist[[y]])))
> olMA[1:12,]
```

| | S1 | S2 | S3 | S4 | S5 | S6 | S7 | S8 | S9 | S10 | S11 | S12 | S13 | S14 | S15 | S16 | S17 | S18 | S19 | S20 |
|-----|----|----|----|----|----|----|----|----|----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|
| S1 | 10 | 4 | 2 | 2 | 5 | 6 | 7 | 6 | 6 | 7 | 6 | 5 | 6 | 7 | 7 | 8 | 7 | 5 | 9 | 6 |
| S2 | 4 | 11 | 5 | 5 | 5 | 6 | 6 | 5 | 6 | 7 | 7 | 7 | 7 | 8 | 7 | 6 | 8 | 6 | 8 | 5 |
| S3 | 2 | 5 | 8 | 2 | 4 | 3 | 5 | 5 | 5 | 7 | 6 | 6 | 6 | 5 | 6 | 5 | 5 | 5 | 6 | 7 |
| S4 | 2 | 5 | 2 | 11 | 6 | 4 | 4 | 2 | 6 | 8 | 7 | 6 | 7 | 8 | 9 | 7 | 5 | 8 | 6 | 5 |
| S5 | 5 | 5 | 4 | 6 | 11 | 6 | 6 | 4 | 6 | 10 | 8 | 6 | 6 | 6 | 9 | 6 | 9 | 9 | 6 | 5 |
| S6 | 6 | 6 | 3 | 4 | 6 | 12 | 6 | 8 | 5 | 8 | 7 | 3 | 7 | 6 | 6 | 6 | 10 | 5 | 6 | 7 |
| S7 | 7 | 6 | 5 | 4 | 6 | 6 | 13 | 8 | 6 | 7 | 9 | 7 | 8 | 7 | 9 | 10 | 8 | 9 | 8 | 8 |
| S8 | 6 | 5 | 5 | 2 | 4 | 8 | 8 | 13 | 7 | 8 | 7 | 6 | 9 | 6 | 9 | 8 | 9 | 7 | 9 | 9 |
| S9 | 6 | 6 | 5 | 6 | 6 | 5 | 6 | 7 | 12 | 9 | 9 | 8 | 9 | 8 | 9 | 8 | 9 | 7 | 9 | 6 |
| S10 | 7 | 7 | 7 | 8 | 10 | 8 | 7 | 8 | 9 | 17 | 11 | 12 | 10 | 13 | 13 | 11 | 11 | 10 | 12 | 10 |
| S11 | 6 | 7 | 6 | 7 | 8 | 7 | 9 | 7 | 9 | 11 | 15 | 10 | 10 | 10 | 11 | 9 | 11 | 11 | 9 | 10 |
| S12 | 5 | 7 | 6 | 6 | 6 | 3 | 7 | 6 | 8 | 12 | 10 | 14 | 9 | 12 | 12 | 10 | 8 | 8 | 12 | 9 |

Exercise 3: nested loops to generate similarity matrices

Task 3.3: Plot the resulting intersect matrix as heat map. The `heatmap.2` function from the `gplots` library can be used for this.

```
> library("gplots")  
> heatmap.2(olMA, trace="none", Colv="none", Rowv="none", dendrogram="none",  
+           col=colorpanel(40, "darkred", "orange", "yellow"))
```



Exercise 4: build your own R package

Task 4.1: Save one or more of your functions to a file called `script.R` and build the package with the `package.skeleton` function.

```
> package.skeleton(name="mypackage", code_files=c("script1.R"), namespace=TRUE)
```

Task 4.2: Build tarball of the package

```
R CMD build mypackage
```

Task 4.3: Install and use package

```
> install.packages("mypackage_1.0.tar.gz", repos=NULL, type="source")  
> library(mypackage)  
> ?myMAcomp # Opens help for function defined by mypackage
```

Additional Exercises

See here: [Link](#)

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What Is L^AT_EX?

- Originally developed in the early 1980s by Leslie Lamport.
- L^AT_EX is a document markup language and document preparation system for the TeX typesetting program.
- Developed for mathematicians, statisticians, engineers and computer scientists.
- High quality of typesetting for scientific articles.
- Programmable environment.
- Many efficient cross-referencing facilities for equations, tables, figures, bibliographies, etc.

How Does It Work?

- Write in your favourite text editor, e.g.: Vim or Emacs.
- Install L^AT_EX distribution for your OS
 - Windows: [MiKTeX](#)
 - Linux: [Latex Project Site](#)
 - Mac OS X: [TexShop](#)
- L^AT_EX manuals (very incomplete selection)
 - List of Manuals on [Latex Project Site](#)
 - [The Not So Short Introduction to LaTeX](#)
 - [Getting to grips with LaTeX](#)
- Packages
 - An almost complete list: [Online TeX Catalogue](#)

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Bibtex: The Ultimate Reference Management Tool

- Sample Latex file: `example.tex`
- Convert to PDF with command: `pdflatex example.tex`
- To include references from MyBibTex.bib database, the following command sequence needs to be executed:
`pdflatex example.tex; bibtex example; pdflatex example.tex`

Examples of BibTeX Citations

- Citation in parentheses (Grant et al., 2006; Schwacke et al., 2003; Miteva et al., 2006)
- Citation of Grant et al. (2006); Schwacke et al. (2003); Miteva et al. (2006)
- Extended citation (Grant et al., 2006, J Chem Inf Model, 46, 1912-1918)
- Footnote citation with more detail (Grant et al., 2006)¹
- ...

The reference list for these citations appears automatically at a defined position, here the end of the document.

¹(Grant et al., 2006, J Chem Inf Model)

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R's Sweave Function Integrates R with Latex

- 'Sweave' provides a flexible framework for mixing Latex and R code for automatic generation of scientific documents.
- It does this by identifying R code chunks - starting with `<<>=` and ending with `@` - and replaces them with the corresponding R output in \LaTeX format, e.g. commands, data objects, plots.
- The user organizes the hybrid code in a `*.Rnw` file, while the `Sweave()` function converts this file into a typical Latex `*.tex` file.
- A quick learning exercise:
 - Download sample hybrid file `Sweave-test-1.Rnw`
 - Run in R command `Sweave("Sweave-test-1.Rnw")`
 - Convert generated `Sweave-test-1.tex` to PDF with `pdflatex Sweave-test-1.tex`
- [Sweave User Manual](#)

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Structure of *.Rnw Hybrid File

Latex $\backslash Sexpr\{pi\}$ Latex Latex Latex Latex Latex Latex
Latex Latex Latex Latex Latex Latex Latex Latex Latex
Latex Latex Latex Latex Latex Latex Latex Latex Latex

<<>>=

R R

R R

@

Latex Latex Latex Latex Latex Latex Latex Latex Latex
Latex Latex Latex Latex Latex Latex Latex Latex Latex
Latex Latex Latex Latex Latex Latex Latex Latex Latex

<<>>=

R R

R R

@

Convert *.Rnw to *.tex to *.pdf

- 1 Create *.tex file

```
> Sweave("mydoc.Rnw")
```

- 2 Create R source file with code chunks (optional)

```
> Stangle("mydoc.Rnw")
```

- 3 Generate PDF with bibliography

```
> system("pdflatex mydoc.tex; bibtex mydoc; pdflatex mydoc.tex")
```

Code Chunk Options

Important options that can be included in code chunk start tag:

`<<◇>>=`

label: optional name for code chunk.

echo: shows command if set to TRUE. Default is TRUE.

fig: shows plots automatically if set to TRUE. Alternatively, one can use standard R code to save graphics to files and point to them in Latex source code.

eval: if FALSE, the variables and functions in code chunk are not evaluated.

prefix: if TRUE, generated file names have a common prefix.

Printing R Commands and Output

Beamer Slide Setting

```
\begin{frame}[containsverbatim]
```

R Code Chunks

```
> <<echo=TRUE>>=  
> 1:10  
> @
```

Result in PDF

```
> 1:10
```

```
[1] 1 2 3 4 5 6 7 8 9 10
```

Including Tables with xtable()

Code Chunk

```
<<echo=FALSE>>=  
library(xtable)  
xtable(iris[1:4,])  
@
```

Result in PDF

| | Sepal.Length | Sepal.Width | Petal.Length | Petal.Width | Species |
|---|--------------|-------------|--------------|-------------|---------|
| 1 | 5.10 | 3.50 | 1.40 | 0.20 | setosa |
| 2 | 4.90 | 3.00 | 1.40 | 0.20 | setosa |
| 3 | 4.70 | 3.20 | 1.30 | 0.20 | setosa |
| 4 | 4.60 | 3.10 | 1.50 | 0.20 | setosa |

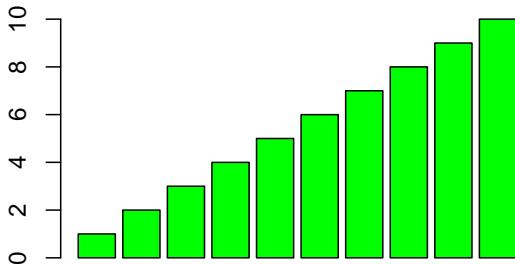
Including Graphics

Code Chunk

```
<<fig=true, width=4.5, height=3.5, eval=TRUE, echo=TRUE>>=  
barplot(1:10, beside=TRUE, col="green")  
@
```

Result in PDF

```
> barplot(1:10, beside=TRUE, col="green")
```



Session Information

```
> sessionInfo()

R version 2.15.0 (2012-03-30)
Platform: x86_64-unknown-linux-gnu (64-bit)

locale:
[1] C

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

other attached packages:
[1] xtable_1.7-0      gplots_2.11.0    MASS_7.3-22
[4] KernSmooth_2.23-8 caTools_1.13     bitops_1.0-5
[7] gdata_2.12.0      gtools_2.7.0

loaded via a namespace (and not attached):
[1] tools_2.15.0
```

Bibliography: to Demo BibTeX I

- Grant, J. A., Haigh, J. A., Pickup, B. T., Nicholls, A., Sayle, R. A., Sep-Oct 2006. Lingos, finite state machines, and fast similarity searching. *J Chem Inf Model* 46 (5), 1912–1918.
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