



R Basics and Differences from Other Languages



Basic Types

Numeric: `x = 2.5`

Integer: `y = as.integer(value)` `v <- c(1L, 2L, 3L)`

Complex: `z = 1 + 2i`

```
> sqrt(-1) Error!
> sqrt(-1 + 0i)
> sqrt(as.complex(-1))
```

Logical: `v = True` `t = 7 > 9`

Character: `name = "Obama"`

Data Structures in R

Vector

(atomic type)

- Collections of cells with elements of the same type
- Default mode is logical

```
> v <- vector()
```

```
> v <- c(#, #, #, ... , #)
```

Matrix

- Vector with an added dimension
- 2 dimensional rectangular layout

```
> m <- matrix(data, nrow, ncol, byrow, dimnames)
```

```
> z <- 1:10
```

```
> dim(z) <- c(2,5)      #2 rows, 5 columns
```

Array

(homogenous type)

- Storing data in **n** dimensions
- Vectors with a dimension attribute

```
> a <- array(1:10, c(2, 5))
```

```
> m <- matrix(1:10, 2)
```

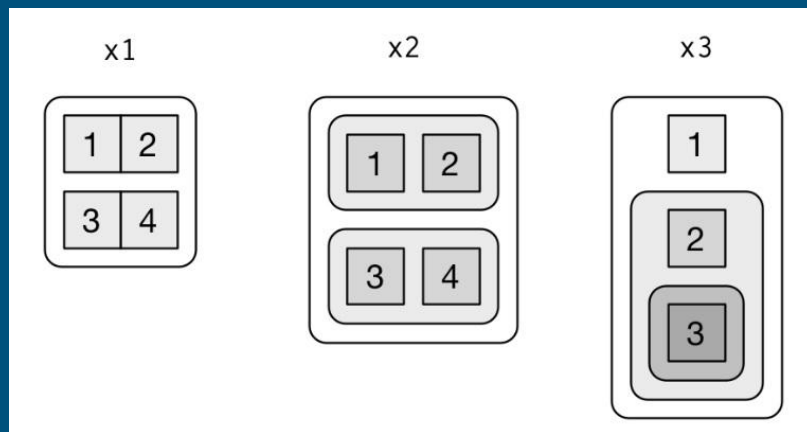
Vector is **not the same** as a 1 dimensional array

Matrix is the **same** as a 2 dimensional array

List

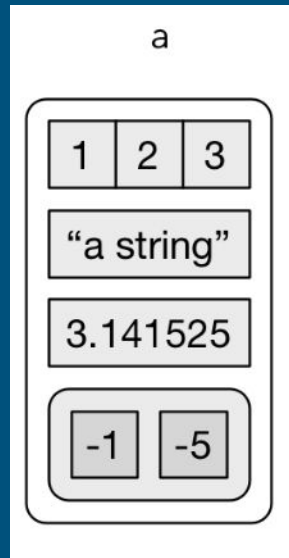
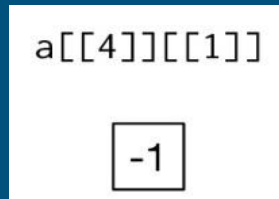
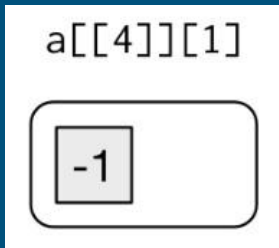
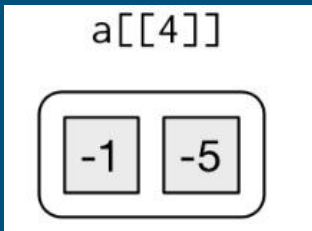
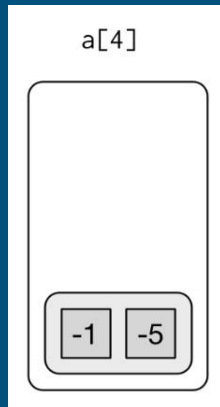
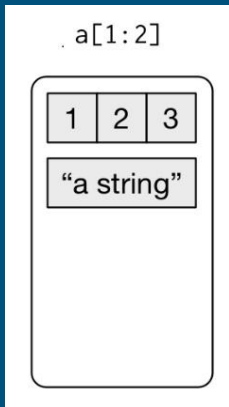
- Mixture of data types
- Can store other lists

```
> x1 <- list(c(1,2), c(3,4))  
> x2 <- list(list(1,2), list(3,4))  
> x3 <- list(1, list(2, list(3)))
```



List exercise

```
> a <- list(1.3, "a string", pi, list(-1, -5))  
> a[1:2]  
> a[4]  
> a[[4]]  
> a[[4]][1]  
> a[[4]][[1]]
```



Dataframe

- Most common way of storing data
- Heterogeneous

```
> firstNames = c("Emma", "Olivia", "Sophia", "Ava", "Isabella", "Mia")  
  
> df <- data.frame(ID = letters[1:6], HOME = sample(state.name, 6), NAME=firstNames)  
  
> df
```

	ID	HOME	NAME
1	a	Illinois	Emma
2	b	Nebraska	Olivia
3	c	Indiana	Sophia
4	d	New Jersey	Ava
5	e	Maryland	Isabella
6	f	Alabama	Mia

Base R & Tidyverse

Base R

built in “out of the box” R

standard object types

indexing based
(e.g. `df[df$group == “a”,]`)

no method chaining

Tidyverse

independent packages

dataframes and tibbles

verb based
(e.g. `filter(df, group == “a”)`)

Lots of method chaining (`%>%`)

R vs. Python



R & Python



R & Python

R

1. R is an implementation of S programming language (Bell Labs).

R focuses on better, user friendly data analysis, statistics and graphical models.

R is usable for basic data analysis without the installation of packages. Big datasets require the use of packages such as `data.table` and `dplyr`, though.

Python

1. Python was inspired by C, Modula-3, and particularly ABC.

Python emphasizes productivity and code readability.

You need to use **NumPy** and **pandas** (amongst others) to make Python usable for data analysis.



R & Python

R Is Slow



R is slow, on purpose



R was designed to make data analysis and statistics easier to do, not to make life easier for your computer.

R has an incomplete informal definition; It is mostly defined in terms of how its implementation works.

Beyond design and implementation, a lot of R code is slow simply because it's poorly written.

Packages to improve R's performance:

parR	A new version of the R interpreter
renjin, FastR	Original R rewritten in Java
Riposte	A fast interpreter and JIT for R
RevoScaleR	Commercial tool to handle big datasets
Foreach	Commercial tool that facilitates parallel programming

Python And Visualizations

"Visualizations are important criteria in choosing data analysis software"

Python has some nice visualization libraries:

Seaborn	Library based on matplotlib
Bokeh	Interactive visualization library
Pygal	To create dynamic svg charts



But there are a lot of options to choose from; Maybe too many.

Moreover, in comparison to R

"Visualizations in Python are usually more convoluted, and the results are not nearly as pleasing to the eye or as informative."



R & Python

R's Steep Learning Curve	Python Is Immature ("It's a challenger!")						
<p>"The worst thing about R is that ... it was developed by statisticians."</p> <p>R's learning curve is nontrivial:</p> <ul style="list-style-type: none">• Even though anybody can get results using GUIs, none is comprehensive enough to totally avoid programming.• Finding packages can be time consuming <p>Using the right tools</p> <p>Good resources can help you to overcome this steep learning curve:</p> <ul style="list-style-type: none"> DataCamp's interactive exercises and tutorials Rdocumentation to search for packages	<p>A more limited way to think about data analysis</p> <p>At the moment, there are no module replacements for the 100s of essential R packages</p> <p>Python's catching up, but will this make people give up R?</p> <ul style="list-style-type: none">• IPython's R extension allows you to cleanly use R in the IPython notebook.• The current landscape of conventions and resources plays a huge role: <table border="0"><tr><td>Matlab</td><td>Commonly used to publish open research code</td></tr><tr><td>Python</td><td>Used in mathematics</td></tr><tr><td>R</td><td>Used in statistics</td></tr></table> <p>Matlabwrap offers a bridge from Python to Matlab, but there are some drawbacks:</p> <ul style="list-style-type: none">- You need to work with two languages- You need a Matlab license	Matlab	Commonly used to publish open research code	Python	Used in mathematics	R	Used in statistics
Matlab	Commonly used to publish open research code						
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R & Python: Data Structures

R

atomic vector, list

named list

matrix, array

dataframe

Python

list, tuple

dictionary

matrix, array (via numpy)

dataframe (via pandas)