Statistical Computing (36-350)

Lecture 2: More Data Structures

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- Matrices
- Lists
- Data frames
- Structures of structures

Recall from last time

Data has different types or classes

Data structures group related values

Vectors are the basic data structure: sequence of values of the same type

```
> x <- c(8, 7, 10, 45)
> x[1]
[1] 8
> x[-4]
[1] 8 7 10
> x[x9]
[1] 10 45
```

Vector structures are vectors + extra attributes Multi-dimensional arrays are vector structures: access through the fancy multiple indices, or through the underlying vector

Running example: resource allocation (continued)

Factory can make cars or trucks
Each car takes 40 hours of labor and 1 ton of steel
Each truck takes 60 hours of labor and 3 tons of steel
1600 hours and 70 tons per week
How many cars and trucks can it make each week?
Can it make (say) 20 trucks and 10 cars per week?
Motivation: step towards linear (mathematical) programming

Matrices

In R, a matrix is a specialization of a 2D array

```
> factory <- matrix(c(40,1,60,3),nrow=2)
> factory
      [,1] [,2]
[1,]      40      60
[2,]      1      3
> is.array(factory)
[1] TRUE
> is.matrix(factory)
[1] TRUE
```

could also specify ncol, and/or byrow=TRUE to fill by rows.

Element-wise operations with the usual arithmetic and comparison operators (e.g., factory/3)

Compare whole matrices with identical() or all.equal()

Matrix multiplication has a special operator:

Multiplying by a vector:

```
> output <- c(10,20)
> factory %*% output
     [,1]
[1,] 1600
[2,] 70
> output %*% factory
     [,1] [,2]
[1.] 420 660
```

R silently casts the vector as a row or column matrix

Matrix transpose:

Matrix determinant:

```
> det(factory)
[1] 60
```

Extracting or replacing the diagonal:

```
> diag(factory) # What's the diagonal of the matrix?
[1] 40 3
> diag(factory) <- c(35,4) # Change it
> factory # See that it changed
      [,1] [,2]
[1,] 35 60
[2,] 1 4
> diag(factory) <- c(40,3) # Set it back for later</pre>
```

Creating a diagonal matrix or an identity matrix:

```
> diag(c(3,4))
     [,1] [,2]
[1,]     3      0
[2,]     0     4
> diag(2)
     [,1] [,2]
[1,]     1      0
[2,]     0      1
```

Inverting a matrix:

Why is it called solve()? Solving the linear system $\mathbf{A}\vec{x} = \vec{b}$, for unknown vector \vec{x} :

```
> available <- c(1600,70)
> solve(factory,available)
[1] 10 20
> factory %*% solve(factory,available)
       [,1]
[1,] 1600
[2,] 70
```

Names in Matrices

We can name either rows or columns or both, with rownames() and colnames()

These are just character vectors, and we use the same function to get and to set their values

Names help us understand what we're working with

Names can be used to coordinate different objects

Example: someone gets the order of cars and trucks wrong

```
> rownames(factory) <- c("labor", "steel")</pre>
> colnames(factory) <- c("cars","trucks")</pre>
> factory
      cars trucks
labor 40
                60
steel 1
> output <- c(20.10)
> names(output) <- c("trucks", "cars")</pre>
> available <- c(1600.70)
> names(available) <- c("labor", "steel")</pre>
> factory %*% output # But we've got cars and trucks mixed up!
      Γ.17
labor 1400
steel 50
> factory %*% output[colnames(factory)]
      Γ.17
labor 1600
steel 70
> all(factory %*% output[colnames(factory)] <= available[rownames(factory)])</pre>
[1] TRUE
```

Notice: Last lines don't have to change if we add motorcycles as output or rubber and glass as inputs (abstraction again).

Doing the same thing to each row or column

Take the mean: rowMeans(), colMeans(): input is matrix, output is vector. Also rowSums(), etc. summary(): Applies vector-style summary to each column apply(), takes 3 arguments: matrix, then 1 for rows and 2 for columns, the name of the function to apply to each

```
> rowMeans(a.matrix)
[1] 22.5 6.0
> apply(a.matrix,1,mean)
[1] 22.5 6.0
```

Lists

Sequence of values, not necessarily all of the same type

```
> my.distribution <- list("exponential",7,FALSE)
> my.distribution
[[1]]
[1] "exponential"

[[2]]
[1] 7

[[3]]
[1] FALSE
```

Most of things which you can do with vectors you can also do with lists

Accessing bits of lists

```
Access: can use [] as with vectors
or use []], but only with a single index (and it drops names and
structures)
> is.character(my.distribution)
[1] FALSE
> is.character(my.distribution[[1]])
[1] TRUE
> my.distribution[2]^ 2
Error in my.distribution[2]^2: non-numeric argument to binary operator
> my.distribution[[2]]^2
Γ1  49
(What happens when you try []] on a vector?)
```

Add to lists with c() (also works with vectors):

```
> my.distribution <- c(my.distribution,7)
> my.distribution
[[1]]
[1] "exponential"
[[2]]
[1] 7
[[3]]
[1] FALSE
[[4]]
[1] 7
```

Chop off the end of a list by setting length to something smaller (also works with vectors):

```
> length(my.distribution)
[1] 4
> length(my.distribution) <- 3
> my.distribution
[[1]]
[1] "exponential"
[[2]]
[1] 7
[[3]]
[1] FALSE
```

Names in lists

We can name some or all of the elements of a list

```
> names(my.distribution) <- c("family", "mean", "is.symmetric")</pre>
> my.distribution
$family
[1] "exponential"
$mean
Γ17 7
$is.symmetric
[1] FALSE
Then we access by name, using $ (which removes names and
structure):
> my.distribution$family
[1] "exponential"
> my.distribution[["family"]]
[1] "exponential"
> my.distribution["family"]
$family
[1] "exponential"
```

Using names when we make the list:

[1] TRUE

```
> another.distribution <- list(family="gaussian",mean=7,sd=1,is.symmetric=TRUE)
> another.distribution
$family
[1] "gaussian"
$mean
[1] 7
$sd
[1] 1
$is.symmetric
```

or after:

```
> my.distribution$was.estimated <- FALSE
> my.distribution[["last.updated"]] <- "2011-08-30"
> my.distribution
$family
[1] "exponential"
$mean
[1] 7
$is.symmetric
[1] FALSE
$was.estimated
[1] FALSE
$last.updated
[1] "2011-08-30"
```

Remove an entry in the list by assigning it the value NULL; try my.distribution\$was.estimated<-NULL

Key-Value Pairs

Lists give us a way to store and look data up by name rather than number (key-value pairs, dictionary, associative array, hash) If all our distributions have a component named family, we can look it up by name without caring where it is in the list (More abstraction)

Data frames

Data frame = classic data table, with n rows for cases, and p columns for variables

Lots of the really-statistical parts of R presume data frames Not just a matrix because every column can be of a different type A hybrid of a matrix and a list; can access columns either like a

matrix or like named parts of a list

Many functions for matrices also work on data frames (rowSums(), summary(), apply(), ...)

Cannot do matrix multiplication on a data frame even if it's all numbers

Example:

```
> a.matrix <- matrix(c(35.8.10.4).nrow=2)</pre>
> colnames(a.matrix) <- c("v1","v2")</pre>
> a.matrix
     v1 v2
[1,] 35 10
[2,] 8 4
> a.matrix$v1 # The $ access operator doesn't work on a matrix
Error in a.matrix$v1 : $ operator is invalid for atomic vectors
> a.data.frame <- data.frame(a.matrix,logicals=c(TRUE,FALSE))</pre>
> a.data.frame
  v1 v2 logicals
1 35 10
            TRUE
2 8 4 FALSE
> a.data.frame$v1 # But $ does work on a data frame
[1] 35 8
> a.data.frame[,"v1"]
[1] 35 8
> a.data.frame[1,]
  v1 v2 logicals
1 35 10
            TRIIE
> colMeans(a.data.frame)
      v 1
         v2 logicals
    21.5
             7.0
                       0.5
```

We can add rows or columns to an array or data-frame with rbind() and cbind(), but be careful about forced type conversions

Structures of structures

Lists of lists, lists of vectors, lists of lists of vectors...

This **recursion** lets us build arbitrarily complicated data structures from the basic ones

Lots of complicated objects are lists of data structures

Example: Eigenstuff

eigen() finds eigenvalues and eigenvectors of a matrix
return value is a list of a vector (of eigenvalues) and a matrix (of
eigenvectors)

```
> eigen(factory)
$values
[1] 41.556171 1.443829
$vectors
           Γ.17
                      [,2]
[1,] 0.99966383 -0.8412758
[2,] 0.02592747 0.5406062
> class(eigen(factory))
[1] "list"
> str(eigen(factory))
List of 2
 $ values : num [1:2] 41.56 1.44
 $ vectors: num [1:2, 1:2] 0.9997 0.0259 -0.8413 0.5406
```

With complicated objects, you can access parts of parts (of parts. . .):

Summary

- Matrices act like you'd hope they would
- Lists let us combine different types of data
- Data frames are hybrids of matrices and lists, for classic tabular data
- Use names of components to make data more meaningful and control access
- Recursion lets us build complicated structures