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# Module 2 -

## Exploring FCM data in R

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## Module 2: Exploring FCM data in R

### Module Overview:

- Read in FCS files
- Explore data annotation and structure
- Visualize FCM data in R

## R Refresher: a vector and a matrix

Note: some output is not shown to save space. Run the commands to see the output.

```
1 # Variable Assignment
2 x ← 5
3 y ← 10
4 x + y
5
6 # Vectors (one dimensional)
7 x ← c(1, 4, 5)
8 y ← seq(from = 1, to = 3, by = 1)
9 y
10 x + y
11
12 # Get help on how to construct a matrix (two dimensional)
13 ?matrix
14 A ← matrix(c(1, 2, 3, 11, 12, 13), nrow = 2, ncol = 3,
15           byrow = TRUE)
16 A
17 rbind(x, y)
18 A + rbind(x, y) # see ?rbind for explanation
19 A[1, 3] # Access entry in first row, third column
20 A[2, ] # Access all entries in second row
21 A[, 1] # All entries in first column
```

## R Refresher: A list with named entries

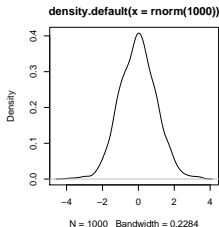
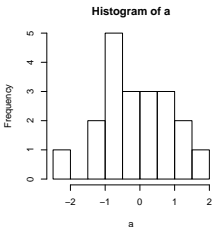
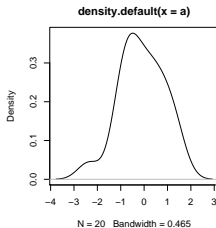
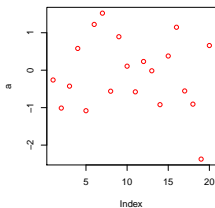
```
1 # Lists and names
2 mylist ← list('first' = x, 'second' = y)
3
4 # Extract the first entry in the list by index (1) or name (
   "first"):
5 mylist[[1]]
6 [1] 1 4 5
7 mylist[["first"]]
8 [1] 1 4 5
9
10 # Checking the size of objects
11 # Vectors and lists have 'length's, matrices have 'dim'
   ensions
12 length(mylist)
13 [1] 2
14 length(x)
15 [1] 3
16 dim(A)
17 [1] 2 3
18 length(A)
19 [1] 6
```

## R Refresher: which, intersect, union

```
1 # Generate 20 random numbers from a normal distribution
2 ?rnorm
3 a ← rnorm(20)
4 a
[1] 0.16215786 0.42892603 -0.40585986 -0.41403764 1.08375740 -0.16635696 -0.11281127
-1.97296631 -1.43725793 -0.61669020 0.92111531 -1.76906195 -0.23469280 -0.28725082
-1.68470666 0.86727154 1.75077450 0.08697872 -1.47718350 -0.03710011

1 # Identify the indices of 'a' > 0, then values < -1
2 which(a > 0)
3 [1] 1 2 5 11 16 17 18
4 which(a < -1)
5 [1] 8 9 12 15 19
6
7 # Check that their intersection is empty, take union
8 intersect(which(a > 0), which(a < -1))
9 [1] integer(0) # empty intersection!
10 combined ← union(which(a > 0), which(a < -1))
11 combined
12 [1] 1 2 5 11 16 17 18 8 9 12 15 19
```

# R Refresher: dot, density, histogram plot



```
1 # Simple plotting functionality
2 plot(a, col = "red")
3 plot(density(a))
4 hist(a)
5 plot(density(rnorm(1000)))
```



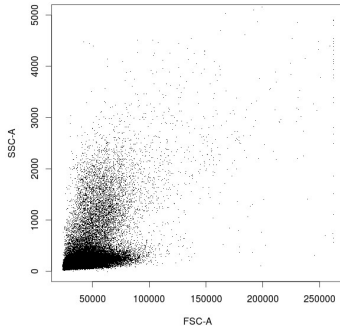
## FCM data: reading FCS files in R

```
1 # Load the package for reading FCM data
2 library(flowCore)
3 # Make sure R knows which directory our data is in
4 setwd('/home/rguru/Documents/Workshop/data')
5 f ← read.FCS('fullFCS/100715.fcs')
6 f
flowFrame object '0d8e743a-05fe-4e8b-9ec4-25993c124ee2'
with 65016 cells and 16 observables:
  name      desc  range  minRange maxRange
$P1  FSC-A      262207 23406.000000 262206
$P2  FSC-H      262207 27008.500000 262206
$P3  SSC-A      261588  -8.014621 261587
$P4  B515-A     KI167 261588  -67.282539 261587
$P5  R780-A      CD3 261588  -67.119034 261587
$P6  R710-A      CD28 261588  -44.558552 261587
$P7  R660-A     CD45R0 261588  -79.819801 261587
$P8  V800-A      CD8 261588  -110.409325 261587
$P9  V655-A      CD4 261588  -66.276711 261587
$P10 V585-A     CD57 261588  -110.472748 261587
$P11 V450-A VIVID / CD14 261588  -28.876564 261587
$P12 G780-A     CCR5 261588  -110.527817 261587
$P13 G710-A     CD19 261588  -89.503387 261587
$P14 G660-A     CD27 261588  -51.957710 261587
$P15 G610-A     CCR7 261588  -61.939350 261587
$P16 G560-A     CD127 261588  -33.256630 261587
253 keywords are stored in the 'description' slot
```

## FCM data: exploring FCS meta data

```
1 # 'f' is a flowFrame object which comes with meta data.
2 # The channel labels:
3 colnames(f)
4 # The expression values measured for each cell:
5 E ← exprs(f)
6 dim(E)
7 # Look at first 10 rows (cells) of E
8 E[1:10, ]
9
10 # Study all the 'keyword' information stored in the file
11 f@description
12 # Access each 'keyword' one at a time
13 f@description$'TUBE NAME'
14 # Access parameter information such as range of expression
    values of FSC-A:
15 f@parameters@data
16 f@parameters@data[1, c("minRange", "maxRange")]
17     minRange maxRange
18 $P1      23406    262206
```

# Plotting FCM data in R



```
1 # First load the library which helps plot FCM data
2 library(flowViz)
3 plot(f, c("FSC-A", "SSC-A"), ylim = c(0,5000), smooth=FALSE)
4 # Note SSC-A is the third parameter (P3) and the meta data
   tells us it is to be viewed on a LOG scale:
5 colnames(f)[3]
6 [1] "SSC-A"
7 f@description$'P3DISPLAY'
8 [1] "LOG"
```

## Reading multiple FCS files

```
1 # 'f' above is a flowFrame object.
2 # A set of flowFrame objects is a flowSet object:
3 fs ← read.flowSet(path = 'fullFCS', pattern = ".fcs")
4 fs
5 A flowSet with 3 experiments.
6   column names:
7   FSC-A FSC-H SSC-A B515-A R780-A R710-A R660-A V800-A V655-
8     A V585-A V450-A G780-A G710-A G660-A G610-A G560-A
9
10 # Explore the meta data for the flowSet:
11 sampleNames(fs)
12 [1] "100715.fcs" "109567.fcs" "113548.fcs"
13 length(fs)
14 [1] 3
15
16 # A flowSet object is closely related to a list
17 # You can access the first frame by index or name
18 fs[["100715.fcs"]]
19 fs[[1]]
```

## Using 'fsApply'

```
1 # Use fsApply to get cell counts for all samples
2 nrow(fs[[1]])
3 [1] 65016
4 fsApply(fs, nrow)
5           [,1]
6 100715.fcs 65016
7 109567.fcs 160074
8 113548.fcs 177102
9
10 # Use 'fsApply' to extract the TUBE NAME keyword
11 # This is useful in identifying control and stained, or
12   stimulated and unstimulated samples in large data sets
13 fsApply(fs, function(f) f@description$'TUBE NAME')
14           [,1]
15 100715.fcs "Tube_025"
16 109567.fcs "Tube_017"
17 113548.fcs "Tube_003"
```

LUNCH 1hr