# QUALIFIER: Quality Assessment of Gated Flow Cytometry Data

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#### Abstract

Background The current flowQ package does the quality assessment on the ungated FCM data. However, there is need to identify deviant samples by monitoring the consistencies of the underlying statistical properties of different gated cell populations (such as white blood cells, lymphacytes, monocytes etc). The current package was also not designed for dealing with large datasets. To meet these needs, We developed QUALIFIER package using the gating template created in flow Jo and performing QA checks on different gated populations. It divides the data preprocessing from the actual outlier detection process so that the statistics are calcuated all at once and the outlier detections and visualization can be done more efficiently and interactively. ncdfFlow is used to solve the memory limit issue for large datasets.

keywords Flow cytometry, Quality Assessment, high throughput, svg, flow Workspace, ncdfFlow Set

## 1 Parsing the QA gating template

Optionally, parsing xml workspace can be done in parallel mode. It can speed up the process for large datasets.

- > library(Rmpi)
- > library(multicore)
- > library(ncdfFlow)

ncdfFlow also needs to be loaded in order to support netCDF storage for flow data that will solve the limitation of memory issue,

parseWorkspace function from flowWorkspace package is used to parse the flowJo workspace. If Rmpi and multicore package are loaded and it will automatically switch to the parallel mode and nslaves arugment is used to specify the number of computing nodes used.

- > ws<-openWorkspace("~/QA\_MFI\_RBC\_bounary\_eventsV3.xml")
- > G<-parseWorkspace(ws,execute=TRUE,isNcdf=TRUE,nslaves=6)
- > saveNcdf("G","gatingHierarchy")
- > save(G,file="gatingHierarchy/GS.Rda")

The result G is a GatingSet containing multiple GatingHierarchy within which gated cell populations are stored. Note that this step is most time consuming especially for large datasets. So it is convienient to save the gatingset once the parsing is done so that it be loaded directly from disk later on for the further processing

## 2 Calculating the statistics

This is the second preprocessing step followed by parsing gating template from flow Jo workspace. Firstly, we need to save the gating hierarchies are calculated and the sample annoation data (containing all the meta information about the FCS files and samples) into a global environment.

```
> anno<-read.csv("~/FCS_File_mapping.csv")###read annotation data
> db<-new.env()
> saveToDB(db,G,anno)
```

Then statistics of each gated population is extracted and saved in db.Again,getQAStats can be speeded up by running in parallel mode. It uses parallel package and automatically detection the number of computing nodes available. Optionally nslaves arugment can also be provided to manually specify the computing node.

```
> library(parallel)
> getQAStats(db)
> ls(db)
> db$statsOfGS[1:5,]
```

It is recommend to save all the preprocessed data to avoid the efforts of recomputing from the beginning:

```
> save(db,file="ITN029.rda")#save stats
```

Once this is done, the more interactive quality assessment task can be performed based on the statistics extracted for each gated population.

## 3 Defining qaTasks

We provide a function to create a list of qaTask objects by reading external csv spreadsheet containing descriptions of each QA task:

```
> checkListFile<-file.path(system.file("data",package="QUALIFIER"),"qaCheckList.csv")
> qaTask.list<-makeQaTask(db,checkListFile)
> qaTask.list[1:2]

$MFIOverTime
qaTask: MFIOverTime
Level : Assay
Description : Fluorescence stability over time
Plot type: xyplot
Gated node: MFI
Default formula :MFI ~ RecdDt | channel * stain
<environment: 0x7f8d9963d940>
```

\$NumberOfEvents

qaTask: NumberOfEvents

Level : Tube

Description: Number of Events Collected

Plot type: xyplot Gated node: Total

Default formula :count ~ RecdDt | Tube

<environment: 0x7f8d995ef910>

This is a convenient way to construct multiple qaTasks. Users can also create the individual qaTask by using new method.

## 4 Quality assessment and visualization

The package provides two important methods:qaCheck and plot to perform quality assessment and visualize the QA results. They both use the information stored in qaTask object and the formula, which is given either explicitly by the argument or implicitly by the qaTask object. It is generally of the form  $y \sim x \mid g1 * g2 * ...$ , y is the statistics to be checked in this QA, It must be one of the four types:

"MFI": Median Fluorescence Intensity of the cell population specified by qaTask,

"percent": the percentage of the cell population specified by qaTask in the parent population,

"count": the number of events of the cell population specified by qaTask,

"spike": the variance of intensity over time of each channel ,which indicating the stability of the fluorescence intensity.

x specifies the variable plotted on x-axis (such as date) in plot method.

g1,g2,.... are the conditioning variables, which divide the data into subgroups and apply the outlier detection whitin each individual groups or plot them in different panels. They may also be omitted,in which case the outliers detection is performed in the entire dataset.

For example, RBC Lysis efficiency (percentage of WBC population) check is defined by qaTask .

> gaTask.list[["RBCLysis"]]

qaTask: RBCLysis
Level : Tube

Description : Sufficient RBC lysis

Plot type: xyplot Gated node: WBC\_perct

Default formula :percent ~ RecdDt | Tube

<environment: 0x7f8d994e1050>

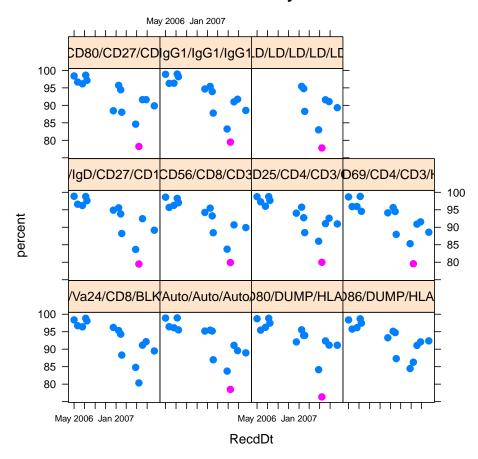
According to the formula stored in qaTask, it uses the statistical property "percent" and groups the data by "Tube" (or staining panel). "RecdDt" is reserved for plotting purpose. Cell population is defined as "WBC-perct"

#### > qaCheck(qaTask.list[["RBCLysis"]],outlierfunc=outlier.cutoff,lBound=0.8)

As we see,qaCheck reads all the necessary information about the gated data from qaTask object. The only thing needs to be specified by end users is how the outliers are called. This is done by providing an outlier detection function outlierfunc that takes a numeric vector as input and returns a logical vector as the output. Here "outlier.cutoff" is used and threshold "lBound" ("less than", using uBound for "larger than") is specified.

#### > plot(qaTask.list[["RBCLysis"]])

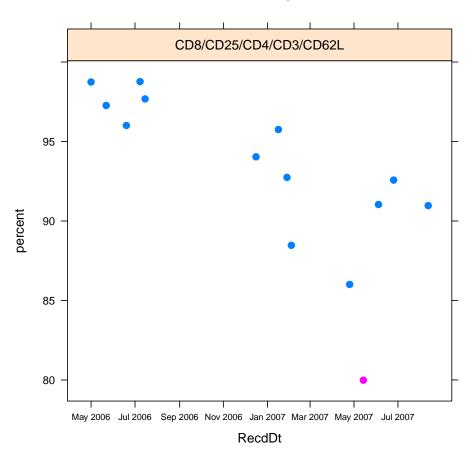
### **Sufficient RBC lysis:**



By default all the data are plotted, argument "subset" can be used to visualize a small subset.

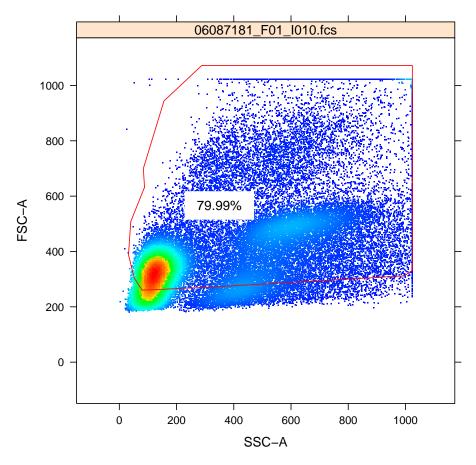
> plot(qaTask.list[["RBCLysis"]],subset="Tube=='CD8/CD25/CD4/CD3/CD62L'")

## **Sufficient RBC lysis:**



With scatterPlot flag set as true and subset properly specified plot method can generate scatter plots for the selected FCS files,

> plot(qaTask.list[["RBCLysis"]],subset="name=='06087181\_F01\_I010.fcs'",scatterPlot=T)



x term in the formula is normally ignored in qaCheck. However, when "plotType" of the qaTask is "bwplot", it is used as the conditioning variable that divides the data into subgroups within which the outlierfunc is applied.

#### > qaTask.list[["MNC"]]

qaTask: MNC
Level : Assay

Description : Consistency of Lymphocyte/MNC Gate

Plot type: bwplot Gated node: MNC

Default formula :percent ~ coresampleid

<environment: 0x7f8d9959cd90>

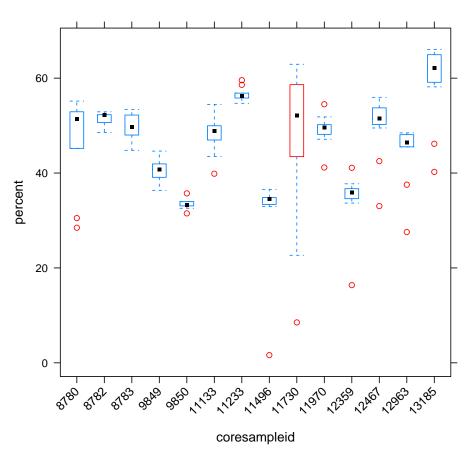
This qaTask detects the significant variance of MNC cell population percentage among aliquots, which have the same "coresampleid". Plot type of this object tells the method to group data by "coresampleid".

#### > qaCheck(qaTask.list[["MNC"]],z.cutoff=1.5)

Interquartile Range based outlier detection function is used to detect outliers

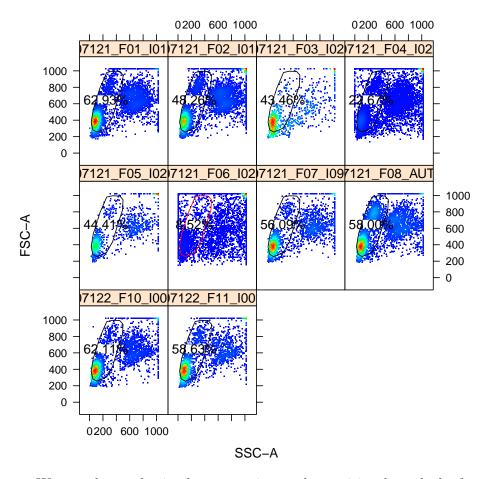
#### > plot(qaTask.list[["MNC"]])

## Consistency of Lymphocyte/MNC Gate:



The red circles in the boxplot indicate the possible outlier samples and the box of red color indicates the entire sample group has significant variance and is marked as the group outlier.

Again, with scatterPlot and subset arguments, scatter plots can be generated for the selected FCS files or sample groups,



We can also apply simple aggregation to the statistics through the formula.

#### > qaTask.list[["BoundaryEvents"]]

qaTask: BoundaryEvents

Level : Channel

Description : Off-scale Boundary Events

Plot type: xyplot Gated node: margin

Default formula :percent ~ RecdDt | channel

<environment: 0x7f8d99445388>

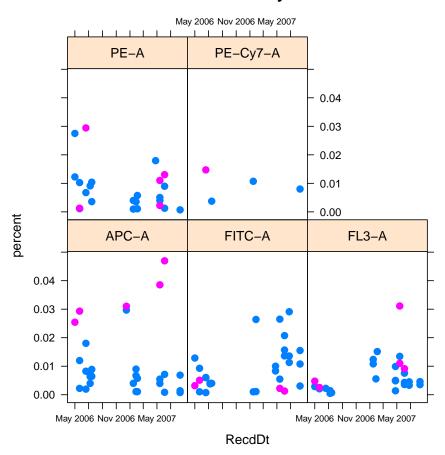
Here the default formula only extracts the "percent" from each individual channel. In order to check the total percentage of boundary events of all channels for each fcs file, we can write a new formula by applying aggregation function "sum" to "percent" and group the data by fcs file ("name" in this case).

```
+ ,uBound=0.0003
+ )
```

And we still can visualize the results chanel by chanel.

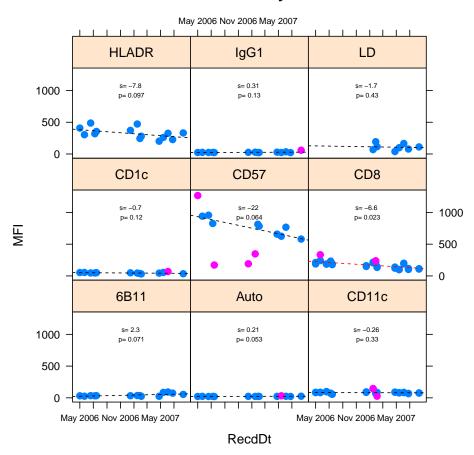
> plot(qaTask.list[["BoundaryEvents"]],percent ~ RecdDt | channel)

#### **Off-scale Boundary Events:**



Another three examples: QA check of Fluorescence stability overtime using t-distribution based outlier detection function.

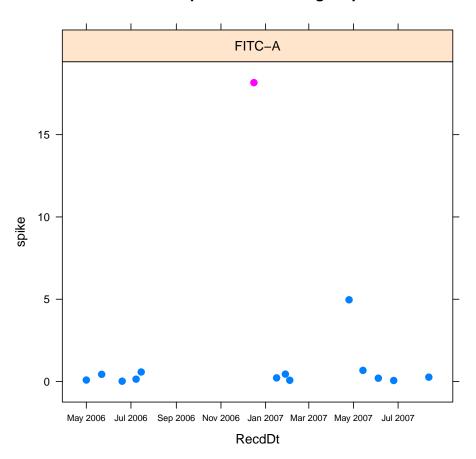
#### Fluorescence stability over time:



Note that the lin-

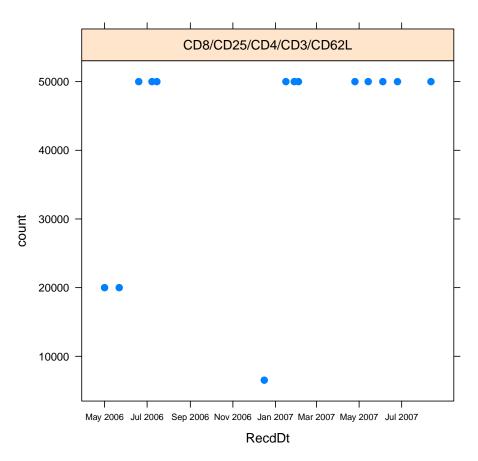
ear regression is applied in each group in order to capture the significant MFI change over time. The individual outliers within each group is detected based on the residue.

## Measurement spikes/drifts during acquisition:



When minitoring the total number of events for each tube, a pre-determined events number can be provided as the threshold to the qaCheck method.

#### **Number of Events Collected:**



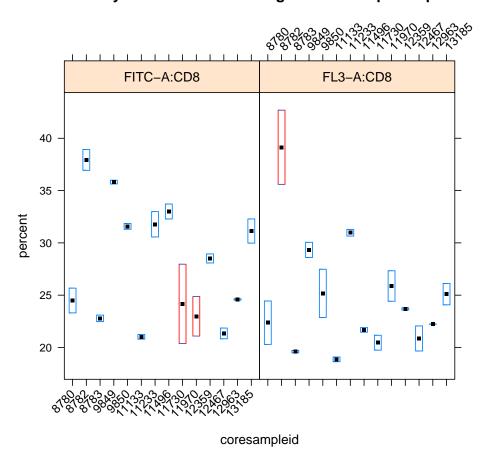
 ${\tt tubesEvents}$  could be a one-column data frame or a named list/vector. Threshold values are stored in the column or list/vector and conditioning values stored in rownames or names of the list/vector.

#### > tubesEvents

	events.EVENTS	events.X2007
CD8/CD25/CD4/CD3/CD62L	250000	50000
CD1c/IgD/CD27/CD19/IgM	250000	150000
6B11/Va24/CD8/BLK/CD4	250000	50000
CD57/CD56/CD8/CD3/CD14	100000	20000
HLADR/CD80/CD27/CD19/CD86	250000	150000
CD8/CD69/CD4/CD3/HLADR	250000	20000
IgG1/IgG1/IgG1/IgG1	20000	20000
Auto/Auto/Auto/Auto	20000	20000
CD11c/CD80/DUMP/HLADr/CD123	400000	400000
CD11c/CD86/DUMP/HLADr/CD123	400000	400000
LD/LD/LD/LD	20000	20000

<sup>&</sup>gt; qaCheck(qaTask.list[["RedundantStain"]],z.cutoff=1)

#### **Consistency of redundant Staining Across sample aliquots:**



5 Creating quality assessment report

Besides the interactive visualization provided by plot method,we also provide one routine to generate all plots in one report. This function reads the QA results calculated by qaCheck and the meta information of each QA task provided in spreadsheet qaCheckList and generate the summary tables and svg plots. Svg plots provide tooltips containing the detail information about each sample as and hyperlinks of densityplot for each individual FCS file.

```
> qaReport(qaTask.list,outDir="~/output",plotAll=FALSE)
```

plotAll is the argument to control the plotting of the individual scatter plot for each FCS file. When TRUE, all the FCS files are plotted. If FALSE, only the FCSs marked as Outliers will be plotted. It can also be set to "none" meaning that no scatter plot will be generated, which provides the quick review of the html report.

## 6 Conclusion

By the formula-based qaCheck and plot methods, different QA tasks can be defined and performed in a generic way. And plot only reads the outliers detection results pre-calculated by qaCheck, which reduces the cost of interactive visualization.

Two kinds of lattice plots are currently supported:xyplot and bwplot(boxplot),depends on the plotType in qaTask object. When the output path is provided by dest, the svg plot is generated. In svg plot, each dot or box (or only the one marked as outliers) is annotated by the tooltip or hyperlink.which further points to the individual density plot of the gated population.