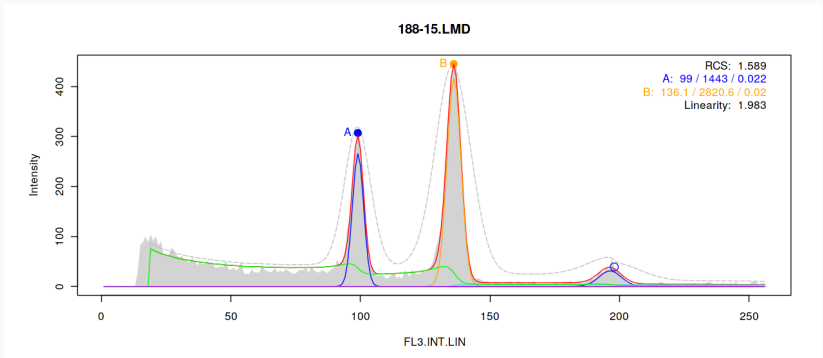


flowPloidy: Determining Genome Size and Ploidy from Flow Cytometry Histograms in R



Tyler Smith¹, Paul Kron² and Sara Martin¹

¹Agriculture and Agri-Food Canada, ²University of Guelph

Motivation

Evolutionary biologists working with polyploid taxa

Population screening:

- large sample sizes
- poor tissue quality

Genome size assessment:

- repeat measurements of fresh/greenhouse tissue
- high precision required

Need **high-throughput, high-precision** genome size estimates

Histogram Analysis

Tissue preparation

Chop and stain tissue

- solution contains 1000s of individual nuclei (and debris)

Flow Cytometer

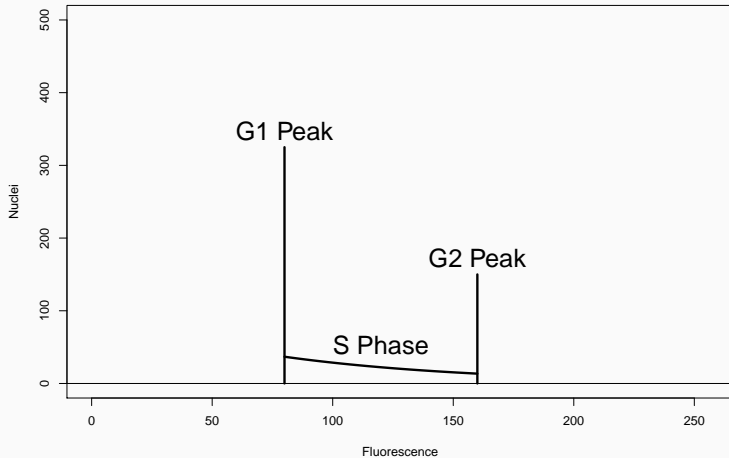
Measure the fluorescence properties of each nuclei:

- size (forward scatter)
- DNA content (fluorescence)
- granularity (side scatter)

Histogram Construction

- aggregate data into bins (256, 512, 1024)

Ideal Histogram



Histogram Components

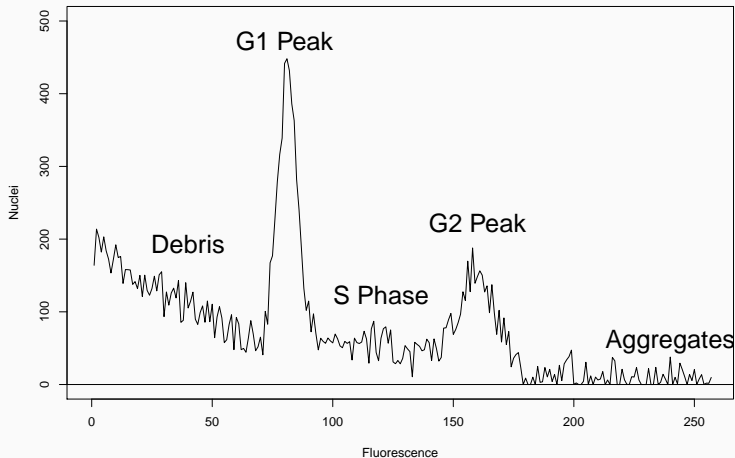
G1 peak Gap 1 diploid cells

G2 peak Gap 2 tetraploid cells (pre-division)

S Phase Synthesis cells (actively duplicating DNA)

(Co-Chopped Standard Peak ignored for now)

Empirical Histogram



Histogram Components

G1 peak Gap 1 diploid cells

G2 peak Gap 2 tetraploid cells (pre-division)

S Phase Synthesis cells (actively duplicating DNA)

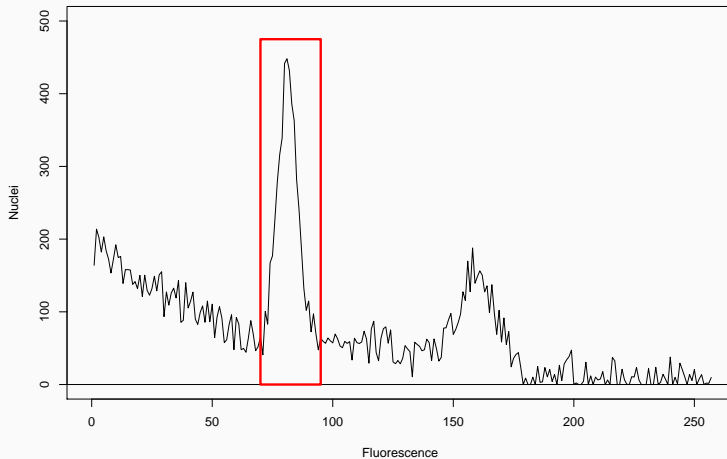
Debris Damaged nuclei, cell components, contaminants

Aggregates clusters of two or more nuclei stuck together

NOISE Measurement error, capriciousness of life

(Co-Chopped Standard Peak ignored for now)

Manual Histogram Analysis



Manual Histogram Analysis

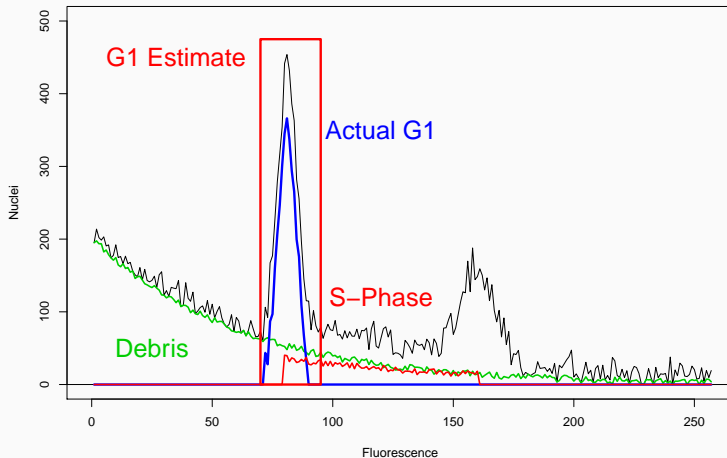
Advantages

- Intuitive
- Several Programs Available
- Can be done ad-hoc in R

Disadvantages

- Subjective
 - CV estimate depends on user
- Doesn't account for overlapping components
 - G1 cell count estimate inflated by debris and S-phase

Overlapping Histogram Components



Non-linear Regression Histogram Analysis

Model histogram components using mathematical functions:

- G1 and G2 peaks fit as Normal curves
- Debris and aggregates fit using theoretical models

Advantages

- objective
- estimates taken directly from the data

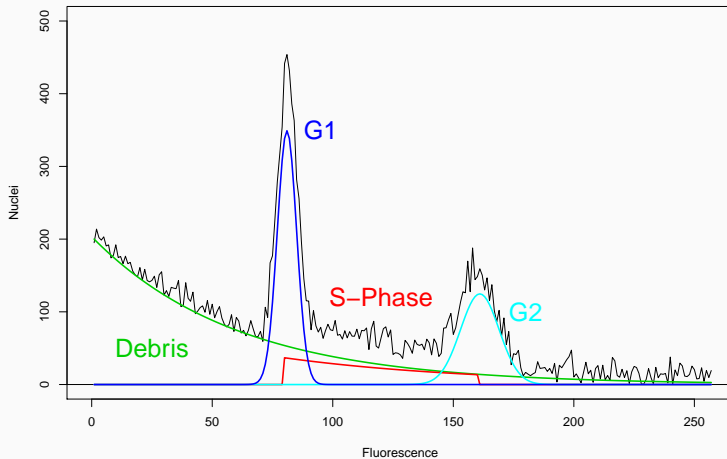
Disadvantages

- availability (few programs, expensive licenses)
- conceptually complex

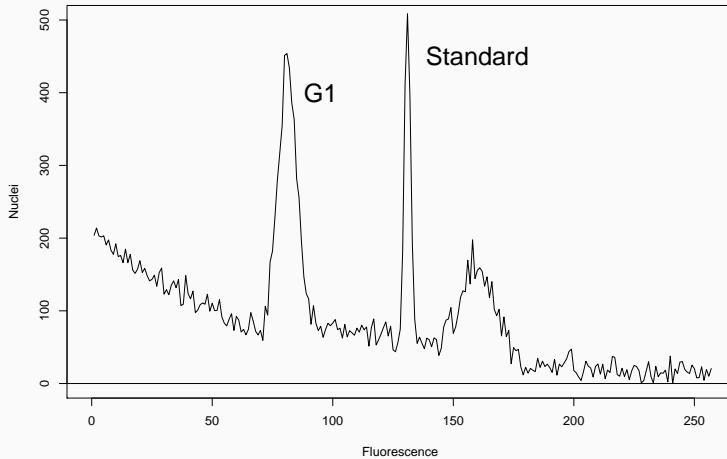
Source

Bagwell, C. B. (1993). Chapt. 3 *In* K. D. Bauer et al., Clinical flow cytometry: principles and applications. Williams & Wilkins.

Non-linear Regression Histogram Analysis



Co-Chopped Standard



Issues with ModFit

- Cost and accessibility
- Functionality (too much and too little)

flowPloidy Goals

- Streamline our workflow, integrate with R
- Increase our understanding of histogram analysis
- Platform for developing best practices

Free Software

Open access for everyone in our lab and yours

You can learn from my work

You can improve and extend my work

Builds on existing tools for flow cytometry in R

Installation is simple:

```
source("https://bioconductor.org/biocLite.R")  
biocLite("flowPloidy")  
biocLite("flowPloidyData")           # for examples
```


Loading Data

```
library(flowPloidy)
library(flowPloidyData) # for examples
batch1 <- batchFlowHist(flowPloidyFiles,
                        channel = "FL3.INT.LIN")

batch1 <- browseFlowHist(batch1)
```

Reviewing Histograms

File 1 of 14

Exit Prev Next

Samples Peak

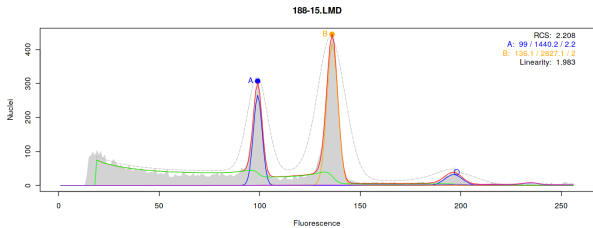
2 A

Standard Value Standard Peak

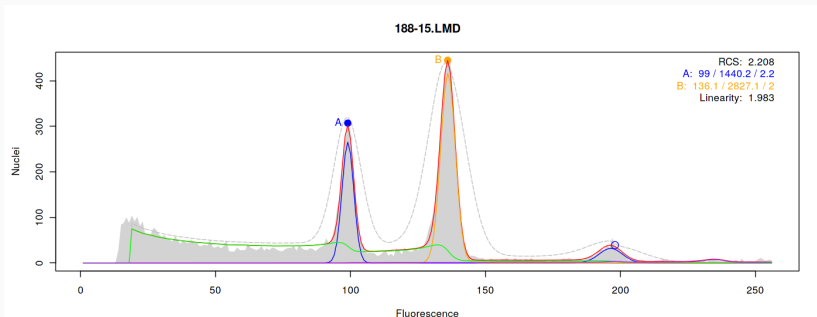
0 X

Linearity Debris Model

Variable SC




Reviewing Histograms



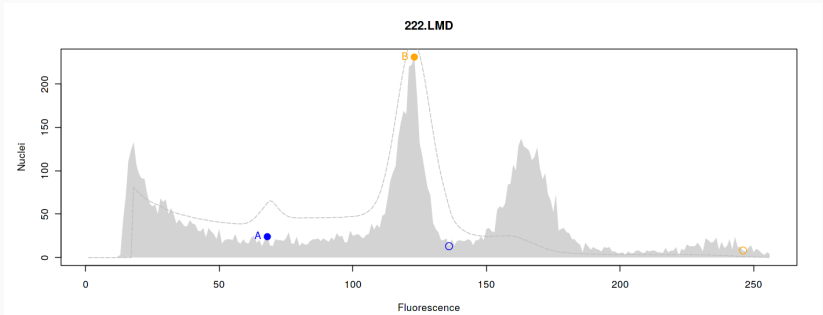
Reviewing Histograms

File 1 of 14

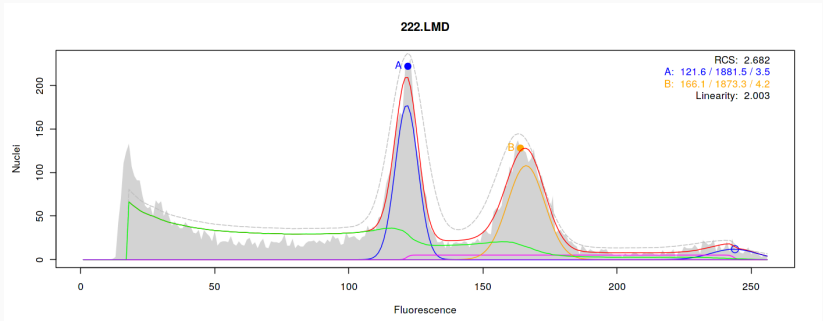
Exit Prev Next

Samples	Peak
2 	A ▼
Standard Value	Standard Peak
0 ▼	X ▼
Linearity	Debris Model
Variable ▼	SC ▼

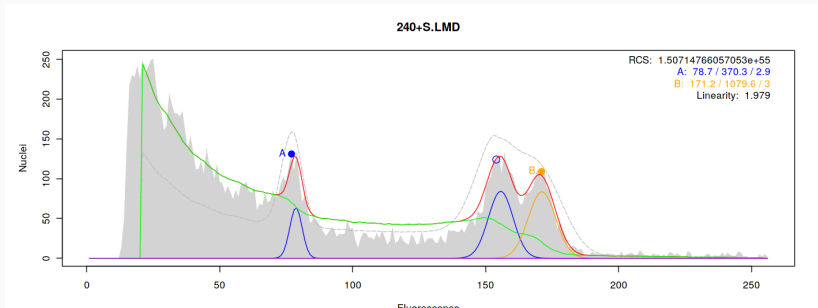
Correcting Histograms



Correcting Histograms




Changing Model Components: Debris



Changing Model Components: Debris

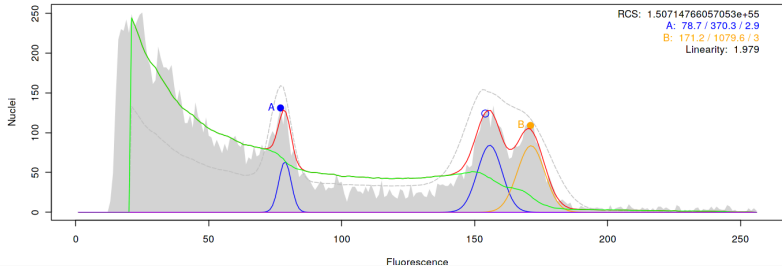
File 1 of 14

Exit Prev Next

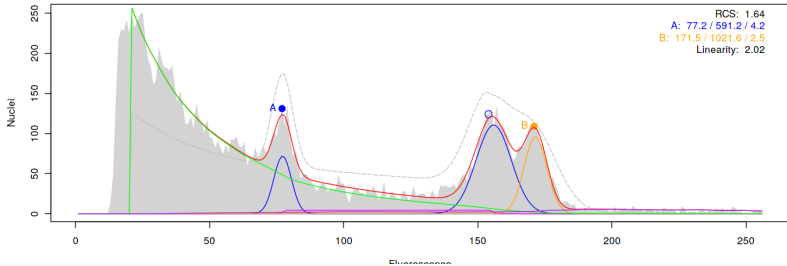
Samples	Peak
2 	A ▼
Standard Value	Standard Peak
0 ▼	X ▼
Linearity	Debris Model
Variable ▼	SC ▼

Changing Model Components: Debris

240+S.LMD



240+S.LMD



Changing Model Components: Debris

One size doesn't fit all

Depends on:

- species
- preparation method
- individual sample quality

Response

Make switching components quick and easy

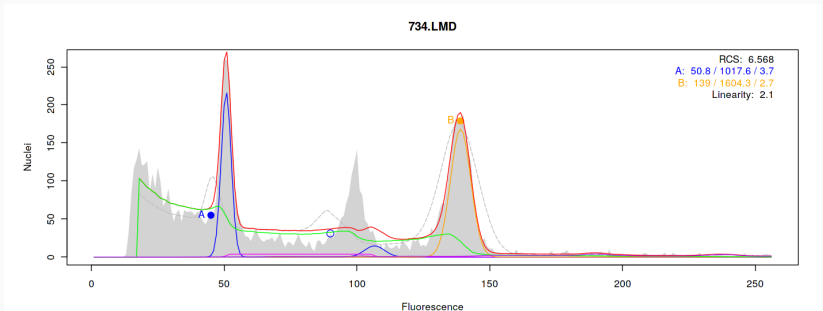
Impact on parameter estimates is usually small

RCS provides an objective basis for choice

Local Minima

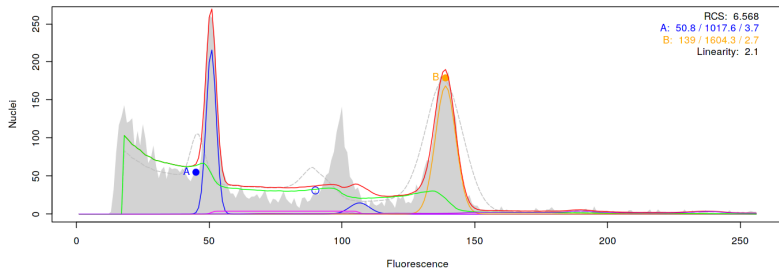
There is no direct solution for non-linear regression

Sometimes the algorithm gets stuck in a local minima:

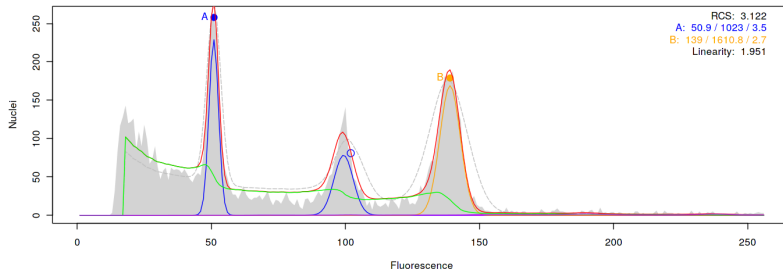


Local Minima

734.LMD



734.LMD



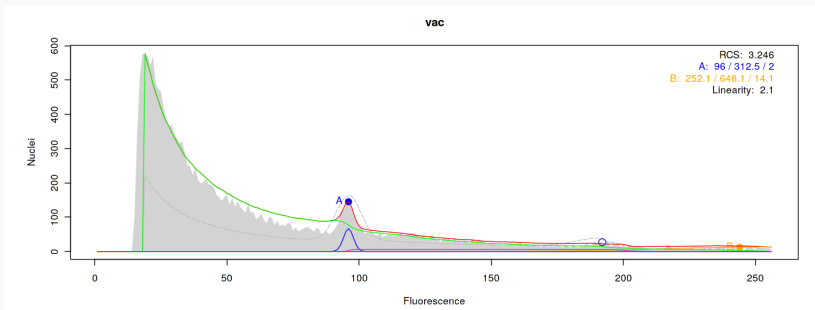
Exporting Results

Save to file or use directly in R:

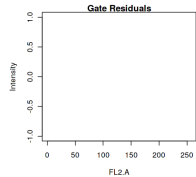
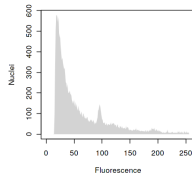
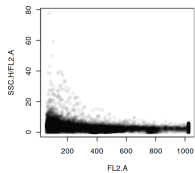
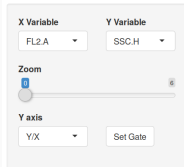
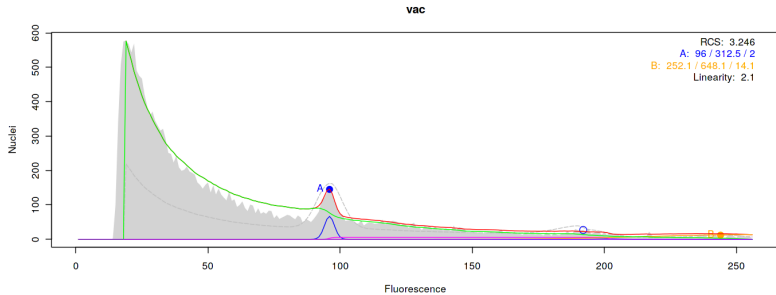
```
tabulateFlowHist(batch1)
```

	countsA	sizeA	cvA	AB
188-15.LMD	1440.229	99.034	0.022	0.727
240-4-2+rad.LMD	449.525	64.598	0.028	0.597
248+S.LMD	2651.879	77.773	0.027	0.395

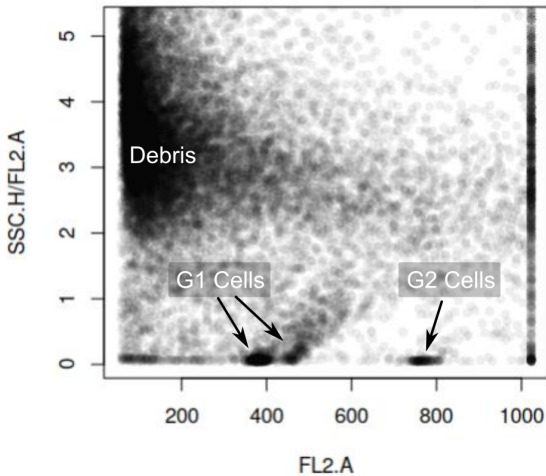
Flow data is not always pretty



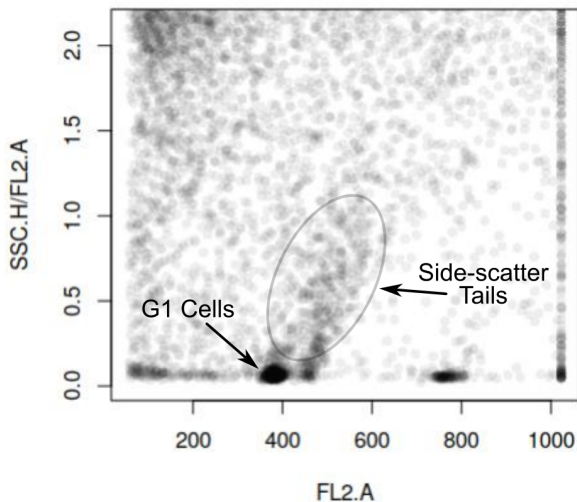
Gating



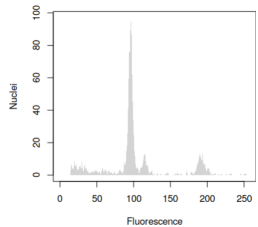
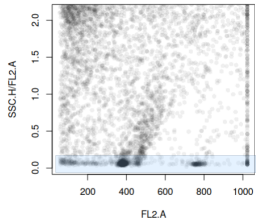
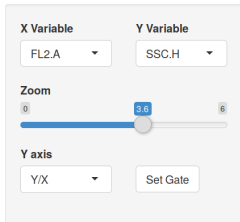
Gating



Gating



Gating



Gating

File 1 of 1

Exit Prev Next

Samples: 2 Peak: B

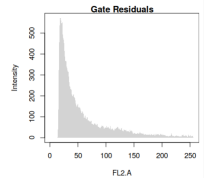
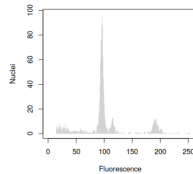
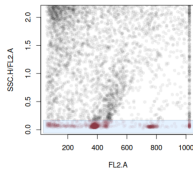
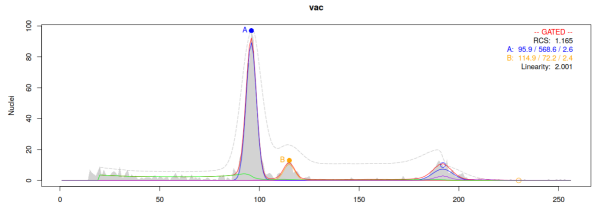
Standard Value: 0 Standard Peak: X

Linearity: Variable Debris Model: SC

X Variable: FL2.A Y Variable: SSC.H

Zoom: 0 3.6 6

Y axis: Y/X Set Gate



flowPloidy Summary

Compared to modFit:

- parameter estimates within 1%
- simpler interface
- integration with R
- cost

Compared to manual analysis:

- objective and repeatable

Gating:

- less sensitive to subjective gating decisions

Gating

- better define or automate best practice
- more sophisticated options for pulse analysis
- impact of gating on theoretical model components

Thank You!

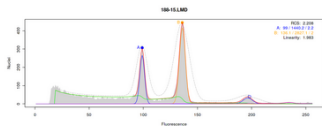
tyler@plantarum.ca

plantarum.github.io/flowPloidy

flowPloidy

An R package for flow cytometry
histogram analysis

[View the Project on GitHub](#)



Introduction

A tutorial overview of `flowPloidy` is available on the [Bioconductor website](#). This vignette is provided with the package, so once you have `flowPloidy` installed you can access it from with R (see below).