A state-of-the-art machine learning pipeline for the analysis of spatial proteomics data

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Introduction

pRoloc and pRolocGUI are R/Bioconductor packages that implement all the necessary tools for the sound and reproducible analysis and interactive exploration of spatial proteomics data from any type of experiment.

Below, we illustrate a typical pRoloc analysis pipeline:

- 1. Loading data into R and adding markers
- 2. QC: checking resolution in the data and organelle markers
- 3. Detection of new organelle clusters
- 4. Classification of unlabelled proteins
- 5. Results, interpretation and visualisation

1) Data input

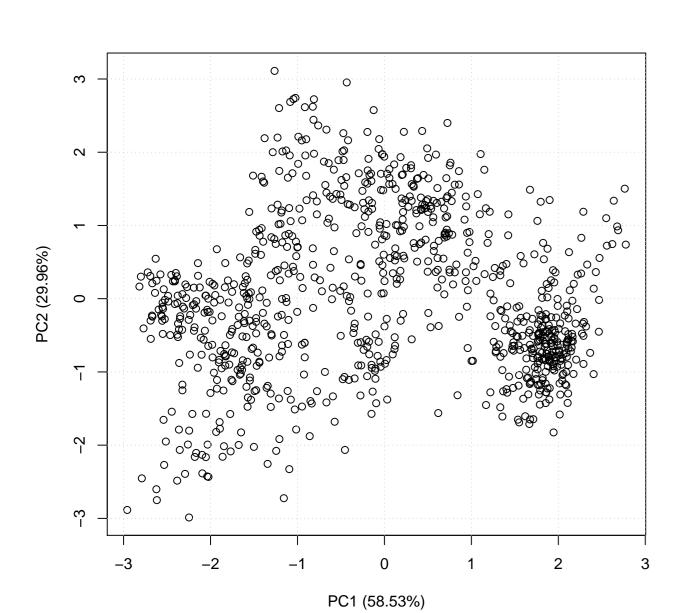
We start by reading quantitative data from 10 fractions (at positions 2 to 11 in the data spreadsheet) sampled along a separation gradient from a csv file and add *Drosophila* organelle markers. This code creates an MSnSet data object (named spat) that stores the quantitative data and the metadata, and can subsequently be easily manipulated, plotted and further processed.

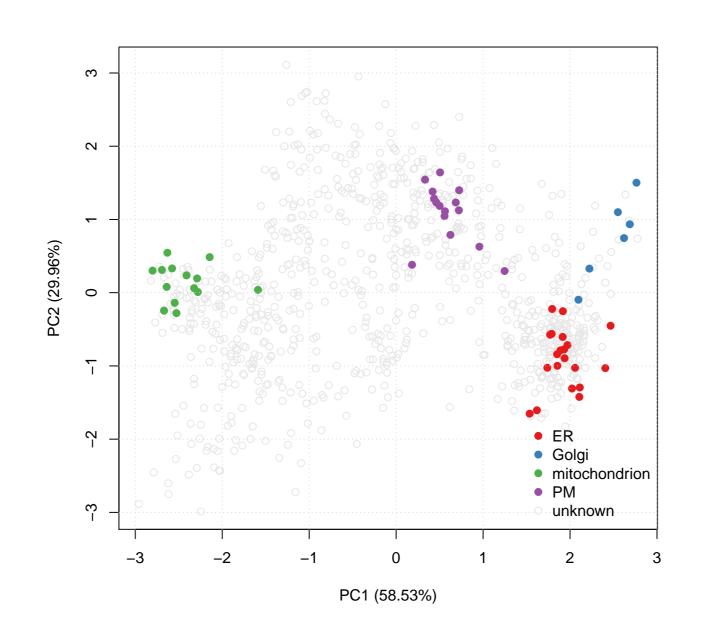
	$Fraction_1$	$\overline{\text{Fraction}_2}$		$Fraction_{m}$		markers	
prot_1	$q_{1,1}$	$q_{1,2}$		q _{1, m}		unknown	
prot_2	$q_{2,1}$	$q_{2,2}$		$q_{2, m}$		$organelle_1$	
prot_3	$q_{3,1}$	$q_{3,2}$		$q_{3, m}$		unknown	
prot_4	$q_{4,1}$	$q_{4,2}$		$q_{4, m}$		$organelle_2$	
•	:	• •	•	:	•	•	
$prot_i$	$q_{i,1}$	$q_{i,2}$		$q_{i, m}$		$organelle_k$	
•	:	• •	• •	:	•	:	
prot_n	$q_{n,1}$	$q_{n,2}$		q _{n, m}	• • •	unknown	
	$Fraction_1$	$Fraction_2$		$Fraction_{m}$			
	• • •	• • •		• • •			
	•	• •	•	: :			
	• • •	• • •					

2) Quality control

We verify on a PCA plot that there is structure in the data (i.e we distinguish well defined clusters, left) and that the markers defined well resolved organelle clusters (right).

plot2D(spat)
addLegend(spat)





Gatto et al. Mass-spectrometry-based spatial proteomics data analysis using pRoloc and pRolocdata. Bioinformatics. 2014 May 1;30(9):1322-4. PMID: 24413670.

Gatto et al. A foundation for reliable spatial proteomics data analysis. Mol Cell Proteomics. 2014 Aug;13(8):1937-52. PMID: 24846987.

software http://is.gd/pRoloc

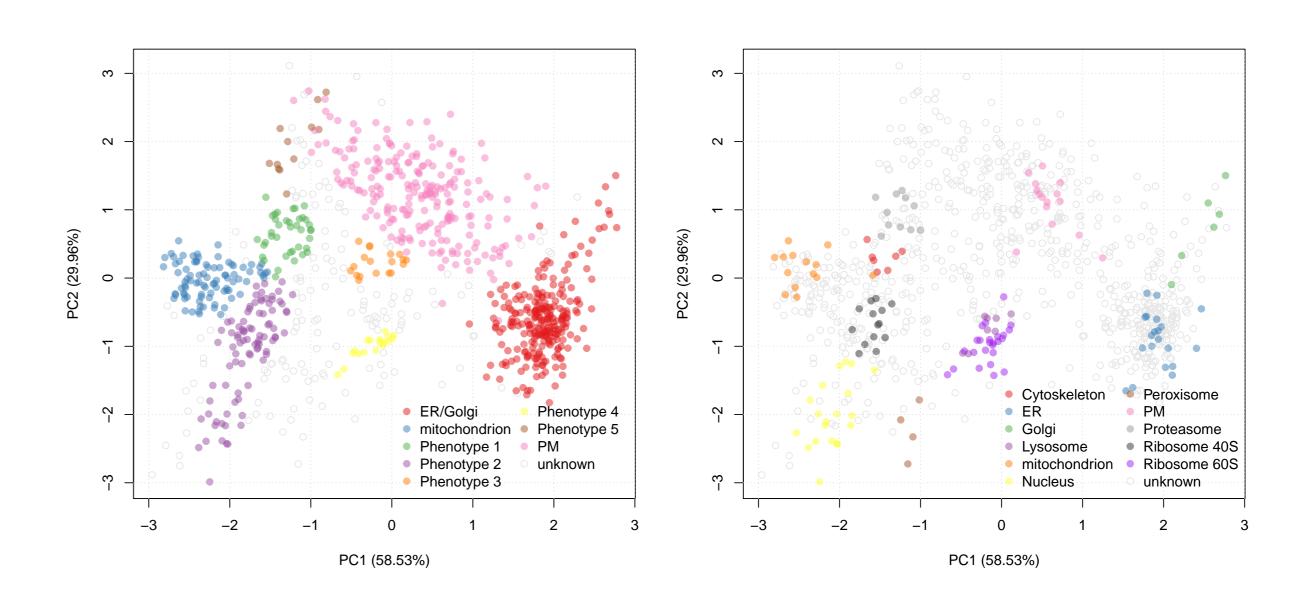
documentation http://is.gd/pRoloc_tutorial

GUIhttp://is.gd/pRolocGUIdatahttp://is.gd/pRolocdataVideosis.gd/R4ProteomicsVideos

3) Novelty detection

Our manually curated markers do not cover the entire sub-cellular diversity. We use a semi-supervised machine learning algorithms (Breckels *et al.*, J Proteomics. 2013 Aug 2;88:129-40.) to identify new putative organelle clusters, called *phenotypes* (left), which require validation by the user (right).

```
spat <- phenoDisco(spat)
plot2D(spat, fcol = "pd")</pre>
```



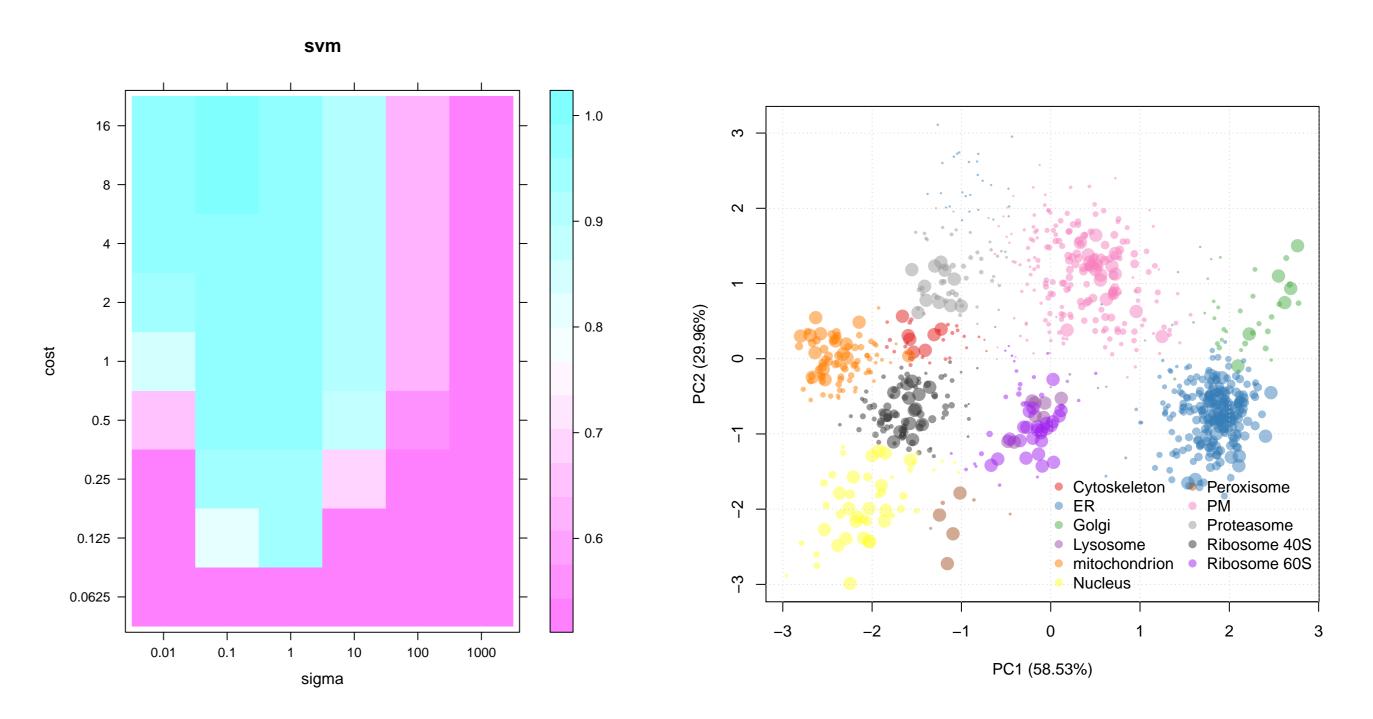
4) Classification

We can now classify unlabelled proteins to any of the augmented classes using a supervised machine (SVM) learning algorithms, for example, a support vector machine classifier. It is essential to tune the classification model parameters (here sigma and cost) prior to actual classification (left).

```
params <- svmOptimisation(spat, fcol = "pd.markers")
spat <- svmClassification(spat, params, fcol = "pd.markers")</pre>
```

The classification algorithm calculates classification probabilities that reflect the distance of a protein to the decision boundaries defined by the SVM model (right). Eventually, the data can be exported to a spreadsheet file.

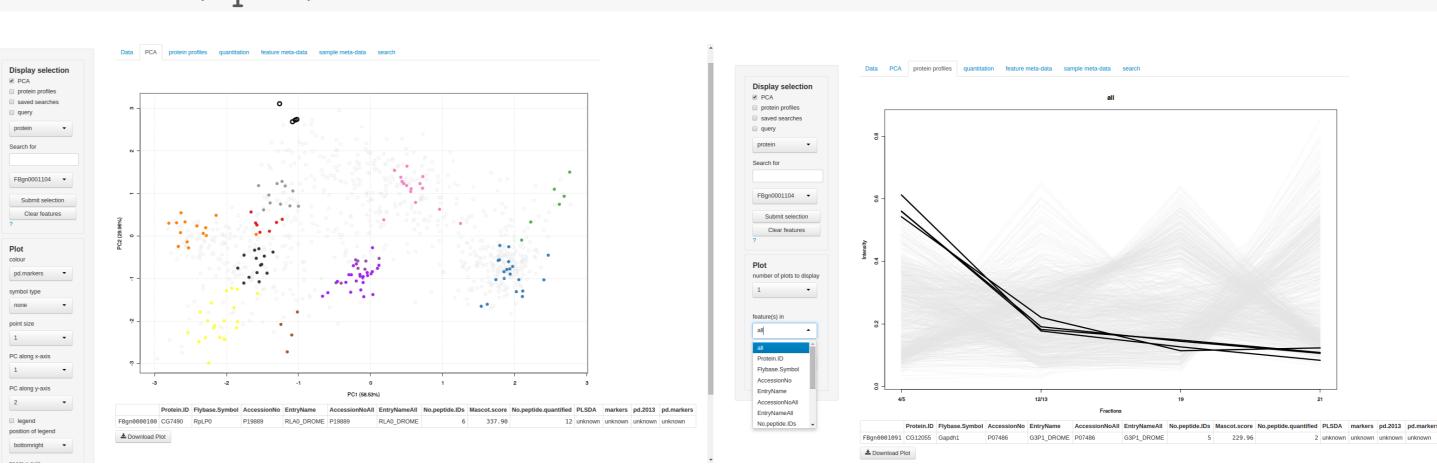
```
ptsze <- exp(fData(spat)$svm.scores) - 1
plot2D(spat, fcol = "svm", cex = ptsze)
write.exprs(spat, file = "spat-results.csv")</pre>
```



5) Interpretation

The graphical user interface implemented in the pRolocGUI package enables one the interactively explore the data.

library("pRolocGUI")
pRolocVis(spat)



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