

Seminar III: R/Bioconductor Splicegear

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Introduction I

- ▶ With the advances in numerical processing of the data, the lowering of the costs and well defined experimental protocols, the reliability of data analysis has increased.
- ▶ However little has been done to quantify how well the technique performs, how the existing oligonucleotide data could have been influenced by alternative phenomenon and how it could contribute to discover novel splice variants.

Introduction I

- ▶ This package defines classes to handle probe expression values in an alternative splicing context.

Installing and loading the package I

```
> library(splicegear)
```

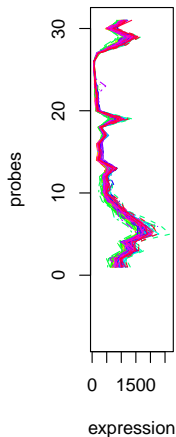
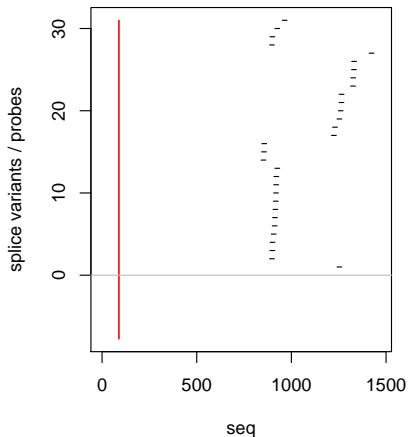
Classes and plotting

The plots are defined for the classes `SPliceSites` and `SpliceExprSet`.

Classes and Plotting I

```
> data(spliceset)
> plot(spliceset)
```

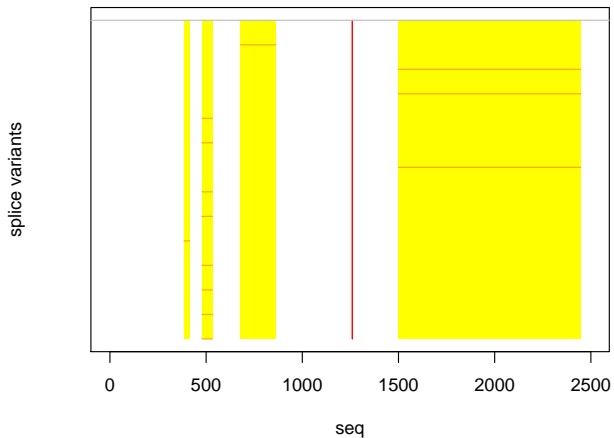
Classes and Plotting II



Classes and Plotting I

```
> data(spsites)
> plot(spsites)
> data(probes)
> plot(probes)
```


Classes and Plotting II



Classes and Plotting I

- ▶ In these data sets you can find information used in the database of putative splice sites against the probes of Aymetrix U95A chips where you can find information about the levels of expression in cells from the liver and central nervous system.

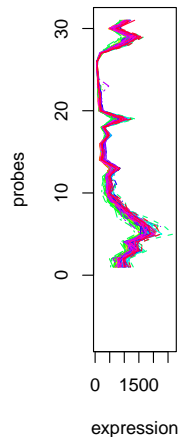
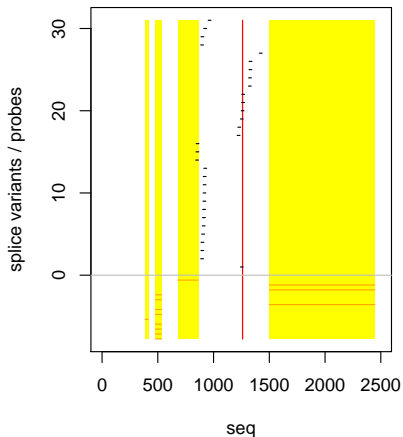
Using the classes I

- ▶ in the first part we described how to plot individual datasets, now let's try one plot a little bit more complicated in which we create a new object of the class `SpliceExprSet` as a result of the fusion of two datasets.

Using the classes I

```
> data(eset, package = "splicegear")  
> data(probes, package = "splicegear")  
> data(spsites, package = "splicegear")  
> spliset <- new("SpliceExprSet",  
+   eset = eset, probes = probes,  
+   spliceSites = spsites)  
> plot(spliset)
```

Using the classes II



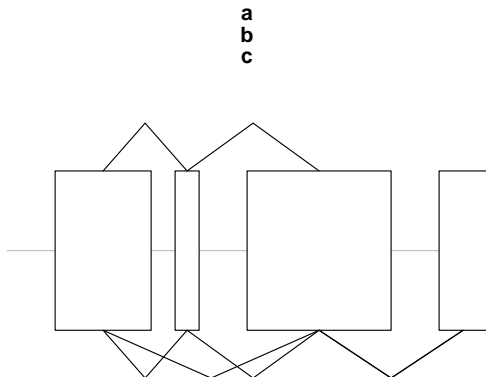
Different splice orders I

- ▶ Now lets create a new sequence in which you can find different splice orders
- ▶ We have to create an object of class SpliceSitesGenomic
- ▶ To make it possible we have to define the possible splice variants, the length of the sequence and the beginning positions of each possible splice site.

Different splice orders I

```
> seq.length <- as.integer(10)
> spsiteIpos <- matrix(c(1, 3.5,
+   5, 9, 3, 4, 8, 10), nc = 2)
> variants <- list(a = c(1, 2, 3,
+   4), b = c(1, 2, 3), c = c(1,
+   3, 4))
> spvar <- new("SpliceSitesGenomic",
+   spsiteIpos = spsiteIpos, variants = variants,
+   seq.length = seq.length)
> plot(spvar)
```

Different splice orders II



Different splice orders I

- ▶ Now to do this a bit more gentle to our eyes lets split the variants apart so you can see each one of them as if there were a plot for each one.

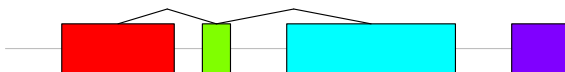
```
> n.exons <- nrow(spsiteIpos)
> par(mfrow = c(3, 1), mar = c(3.1,
+   2.1, 2.1, 1.1))
> plot(spvar, split = TRUE, col.exon = rainbow(n.exons))
```

Different splice orders II

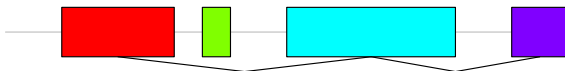
variant a



variant b



variant c



Using an xml view I

- ▶ Now lets import a dataset and predict all its splice sites.

```
> library(XML)
> filename <- system.file("data",
+   "example.xml", package = "splicegear")
> xml <- xmlTreeParse(filename, asTree = TRUE)
> spsites <- buildSpliceSites(xml)
```

7 entrie(s) in the set.

entrie 1 has 23 element(s).

sub-entrie 1 has 1 element(s).

sub-entrie 2 has 2 element(s).

sub-entrie 3 has 3 element(s).

sub-entrie 4 has 1 element(s).

sub-entrie 5 has 3 element(s).

Using an xml view II

```
sub-entrie 6 has 2 element(s).
sub-entrie 7 has 7 element(s).
sub-entrie 8 has 2 element(s).
sub-entrie 9 has 1 element(s).
sub-entrie 10 has 1 element(s).
entrie 2 has 52 element(s).
sub-entrie 1 has 1 element(s).
sub-entrie 2 has 2 element(s).
sub-entrie 3 has 1 element(s).
sub-entrie 4 has 1 element(s).
sub-entrie 5 has 1 element(s).
sub-entrie 6 has 1 element(s).
sub-entrie 7 has 1 element(s).
sub-entrie 8 has 31 element(s).
```

Using an xml view III

```
sub-entrie 9 has 1 element(s).
sub-entrie 10 has 7 element(s).
sub-entrie 11 has 2 element(s).
sub-entrie 12 has 1 element(s).
sub-entrie 13 has 1 element(s).
sub-entrie 14 has 1 element(s).
entrie 3 has 2 element(s).
  sub-entrie 1 has 2 element(s).
entrie 4 has 17 element(s).
  sub-entrie 1 has 2 element(s).
  sub-entrie 2 has 2 element(s).
  sub-entrie 3 has 8 element(s).
  sub-entrie 4 has 1 element(s).
  sub-entrie 5 has 3 element(s).
```

Using an xml view IV

```
sub-entrie 6 has 1 element(s).
entrie 5 has 11 element(s).
sub-entrie 1 has 1 element(s).
sub-entrie 2 has 8 element(s).
sub-entrie 3 has 2 element(s).
entrie 6 has 5 element(s).
sub-entrie 1 has 4 element(s).
sub-entrie 2 has 1 element(s).
entrie 7 has 18 element(s).
sub-entrie 1 has 2 element(s).
sub-entrie 2 has 9 element(s).
sub-entrie 3 has 1 element(s).
sub-entrie 4 has 4 element(s).
```

Using an xml view V

```
sub-entrie 5 has 1 element(s).  
sub-entrie 6 has 1 element(s).
```

Using XML I

```
> data(spliceset)
> dataf <- as.data.frame(spliceset)
> lm.panel <- function(x, y, ...) {
+   points(x, y, ...)
+   p.lm <- lm(y ~ x)
+   abline(p.lm)
+ }
```

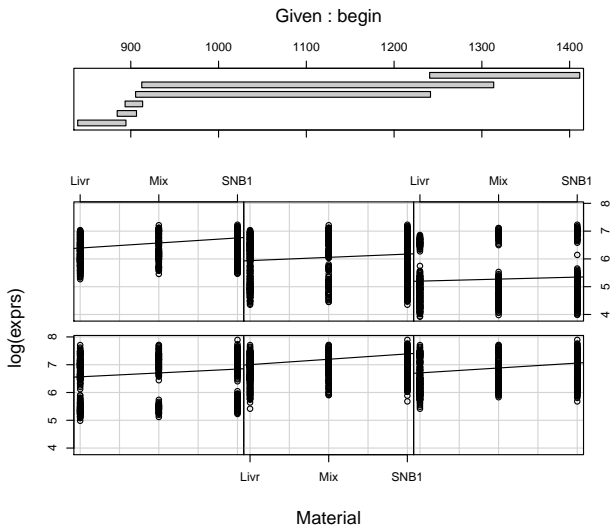

And now? I

- ▶ Probe intensity values conditioned by the position of the probes on the mRNA

One interesting plot I

```
> coplot(log(exprs) ~ Material |  
+       begin, data = dataf, panel = lm.panel)
```

One interesting plot II



The end I

- ▶ Thank you very much!