protViz: Visualizing and Analyzing Mass Spectrometry Related Data in Proteomics

Christian Panse Jonas Grossmann

June 18, 2014

Vignette for package version $\geq v.0.1.48$

Recent changes, updates, and new features:

- 0.1.92 added unit tests.
- 0.1.91 bugfix rt normailzation in `genSwathLib`.
- 0.1.89 replace `fragmentIons` by `fragmentIon`.
- 0.1.77 bugfix xlim in `peakplot`

Contents

1 Preliminary Note

protViz is an R package to do quality checks, visualizations and analysis of mass spectrometry data, coming from proteomics experiments. The package is developed, tested and used at the Functional Genomics Center Zurich. We use this package mainly for prototyping, teaching, and having fun with proteomics data. But it can also be used to do solid data analysis for small scale data sets. Nevertheless, if one is patient, it also handles large data sets.

2 Related Work

The methode of choice in proteomics is mass spectrometry. There are already packages in R which deal with mass spec related data. Some of them are listed here:

- MSnbase package (basic functions for mass spec data including quant aspect with iTRAQ data)
  http://www.bioconductor.org/packages/release/bioc/html/MSnbase.html
- plgem – spectral counting quantification, applicable to MudPIT experiments
  http://www.bioconductor.org/packages/release/bioc/html/plgem.html
- synapter – MSe (Hi3 = Top3 Quantification) for Waters Q-tof data aquired in MSe mode
  http://www.bioconductor.org/packages/release/bioc/html/synapter.html
- mzR
  http://www.bioconductor.org/packages/release/bioc/html/mzR.html
- isobar iTRAQ/TMT quantification package
  http://www.bioconductor.org/packages/release/bioc/html/isobar.html
- readMzXmlData
  http://cran.r-project.org/web/packages/readMzXmlData/
- msQC
  http://bioconductor.org/packages/3.0/bioc/html/msQC.html

3 Get Data In – Preprocessing

The most time consuming and challenging part for data analysis and visualization is shaping the data that they can easily be processed. In this package, we intentionally left this part away because it is very infrastructure dependent. Moreover we use also commercial tools to analyze data and export the data into R accessible formats. We provide different kind of importers if these formats are available, but with very little effort, one can bring other exports in a similar format which will make it easy to use our package for a variety of tools.

3.1 Identification - In-silico from Proteins to Peptides

For demonstration we use a sequence of peptides derived from a tryptics digest using the Swissprot `FETUA_BOVIN Alpha-2-HS-glycoprotein precursor (Fetuin-A) (Asialofetuin)` protein.

`fcat` and `tryptic-digest` are commandline programs which are included in the package. `fcat` removes the lines starting with > and all 'new line' character within the protein sequence while `tryptic-digest` is doing the triptic digest of a protein sequence applying the rule: cleave after arginine (R) and lysine (K) except followed by proline(P).

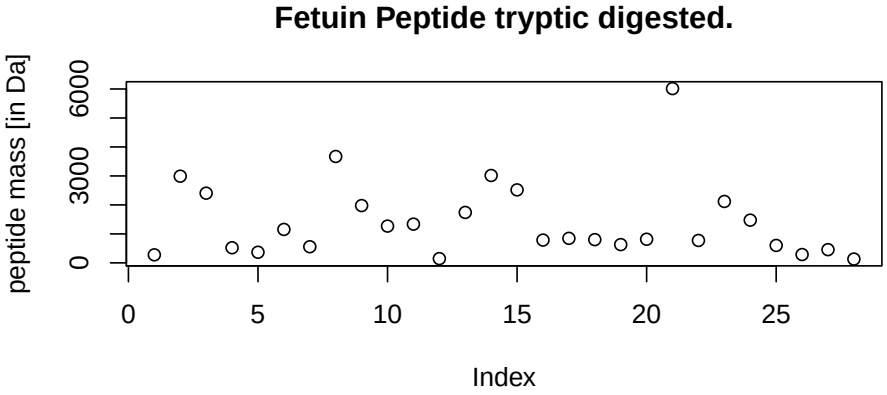
```
$ cat Fetuin.fasta
MKSFVLLFCLAQLWGCHSIPLDPVAGYKEPACDDPDTEQAALAAVDYINKHLPRGYKHTL
NQIDSVKVWPRRPTGEVYDIEIDTLETTCHVLDPTPLANCSVRQQTQHAVEGDCDIHVLK
QDGQFSVLFTKCDSSPDSAEDVRKLCPDCPLLAPLNDSRVVHAVEVALATFNAESNGSYL
QLVEISRAQFVPLPVSVSVEFAVAATDCIAKEVVDPTKCNLLAEKQYGFCKGSVIQKALG
GEDVRVTCTLFQTQPVIPQPQPDGAEAEAPSAVPDAAGPTPSAAGPPVASVVVGPSVVAV
PLPLHRAHYDLRHTFSGVASVESSSGEAFHVGKTPIVGQPSIPGGPVRLCPGRIRYFKI

$ cat Fetuin.fasta | fcat | tryptic-digest
MK
SFVLLFCLAQLWGCHSIPLDPVAGYK
EPACDDPDTEQAALAAVDYINK
HLPR
GYK
HTLNQIDSVK
VWPR
RPTGEVYDIEIDTLETTCHVLDPTPLANCSVR
QQTQHAVEGDCDIHVLK
QDGQFSVLFTK
CDSSPDSAEDVR
K
LCPDCPLLAPLNDSR
VVHAVEVALATFNAESNGSYLQLVEISR
AQFVPLPVSVSVEFAVAATDCIAK
EVVDPTK
CNLLAEK
QYGFCK
GSVIQK
ALGGEDVR
VTCTLFQTQPVIPQPQPDGAEAEAPSAVPDAAGPTPSAAGPPVASVVVGPSVVAVPLPLHR
AHYDLR
HTFSGVASVESSSGEAFHVGK
TPIVGQPSIPGGPVR
LCPGR
IR
YFK
I
```

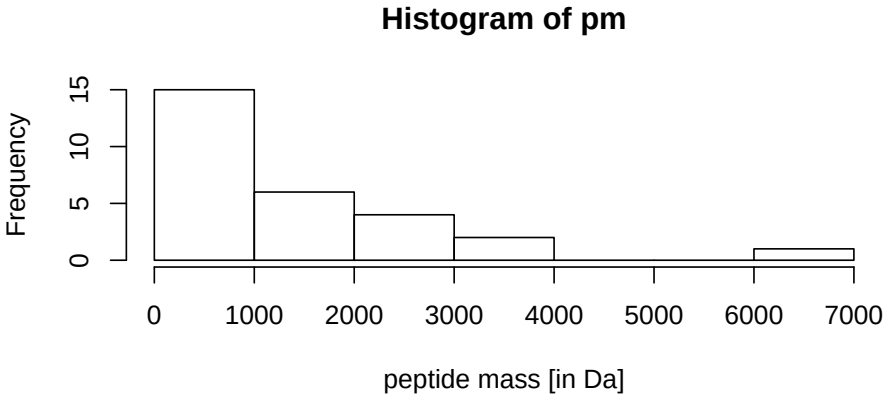
4 Peptide Identification

The currency in proteomics are the peptides. In proteomics, proteins are digested to so-called peptides since peptides are much easier to handle biochemically than proteins. Proteins are very different in nature some are very sticky while others are soluble in aqueous solutions while again are only sitting in membranes. Therefore, proteins are chopped up into peptides because it is fair to assume, that for each protein, there will be a number of peptides behaving well, so that they can actually be measured with the mass spectrometer. This step introduces another problem, the so-called protein inference problem. In this package here, we do not at all touch upon the protein inference.

4.1 Computing the Parent Ion Mass

```
> library(protViz)
> op<-par(mfrow=c(1,1))
> fetuin<-c('MK', 'SFVLLFCLAQLWGCHSIPLDPVAGYK',
+ 'EPACDDPDTEQAALAAVDYINK',
+ 'HLPR', 'GYK', 'HTLNQIDSVK', 'VWPR',
+ 'RPTGEVYDIEIDTLETTCHVLDPTPLANCSVR',
+ 'QQTQHAVEGDCDIHVLK', 'QDGQFSVLFTK',
+ 'CDSSPDSAEDVR', 'K', 'LCPDCPLLAPLNDSR',
+ 'VVHAVEVALATFNAESNGSYLQLVEISR',
+ 'AQFVPLPVSVSVEFAVAATDCIAK',
+ 'EVVDPTK', 'CNLLAEK', 'QYGFCK',
+ 'GSVIQK', 'ALGGEDVR',
+ 'VTCTLFQTQPVIPQPQPDGAEAEAPSAVPDAAGPTPSAAGPPVASVVVGPSVVAVPLPLHR',
+ 'AHYDLR', 'HTFSGVASVESSSGEAFHVGK',
+ 'TPIVGQPSIPGGPVR', 'LCPGR', 'IR', 'YFK', 'I')
> (pm<-parentIonMass(fetuin))

 [1]  278.1533 2991.5259 2406.0765  522.3147  367.1976 1154.6164  557.3194
 [8] 3671.7679 1977.9447 1269.6474 1337.5274  147.1128 1740.8407 3016.5738
[15] 2519.3214  787.4196  847.4342  802.3552  631.3773  816.4210 6015.1323
[22]  774.3893 2120.0043 1474.8376  602.3079  288.2030  457.2445  132.1019

> op<-par(mfrow=c(2,1))
> plot(pm, ylab="peptide mass [in Da]",
+     main="Fetuin Peptide tryptic digested.")
> hist(pm, xlab="peptide mass [in Da]")
```



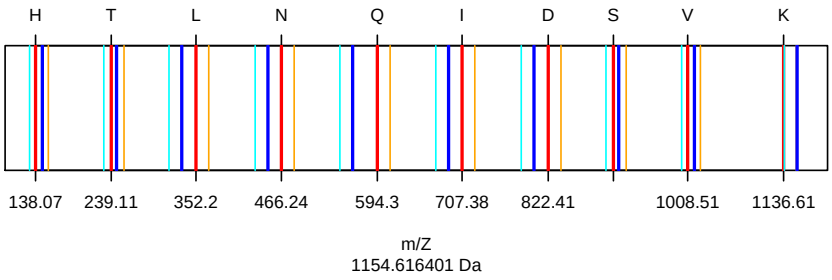


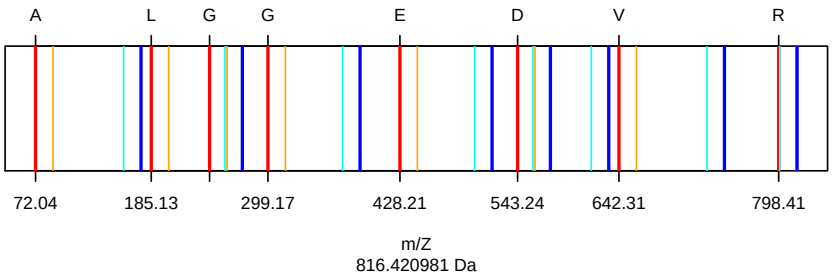
4.2 In-silico Peptide Fragmentation

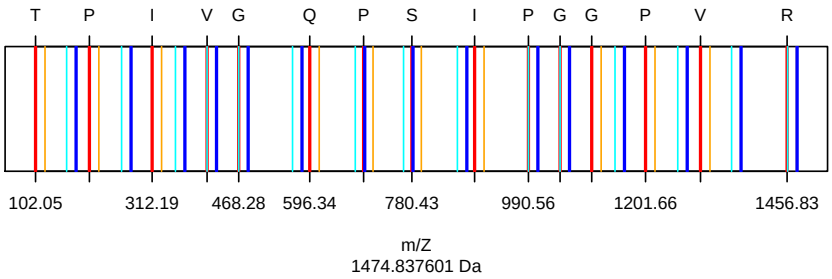
The fragment ions of a peptide can be computed following the rules proposed in [4]. Beside the b and y ions the FUN argument of fragmentIon defines which ions are computed. the default ions beeing computed are defined in the function defaultIon. The are no limits for defining other forms of fragment ions for ETD (c and z ions) CID (b and y ions).

```
> defaultIon

function (b, y)
{
    Hydrogen <- 1.007825
    Oxygen <- 15.994915
    Nitrogen <- 14.003074
    c <- b + (Nitrogen + (3 * Hydrogen))
    z <- y - (Nitrogen + (3 * Hydrogen))
    return(cbind(b, y, c, z))
}
<environment: namespace:protViz>

> peptides<-c('HTLNQIDSVK', 'ALGGEDVR', 'TPIVGQPSIPGGPVR')
> pim<-parentIonMass(peptides)
```

```
> fi<-fragmentIon(peptides)
> par(mfrow=c(3,1));
> for (i in 1:length(peptides)){
+     plot(0,0,
+         xlab='m/Z',
+         ylab='',
+         xlim=range(c(fi[i][[1]]$b,fi[i][[1]]$y)),
+         ylim=c(0,1),
+         type='n',
+         axes=FALSE,
+         sub=paste( pim[i], "Da"));
+     box()
+     axis(1,fi[i][[1]]$b,round(fi[i][[1]]$b,2))
+     pepSeq<-strsplit(peptides[i],"")
+     axis(3,fi[i][[1]]$b,pepSeq[[1]])
+
+     abline(v=fi[i][[1]]$b, col='red',lwd=2)
+     abline(v=fi[i][[1]]$c, col='orange')
+     abline(v=fi[i][[1]]$y, col='blue',lwd=2)
+     abline(v=fi[i][[1]]$z, col='cyan')
+ }
```







The next lines compute the singly and doubly charged fragment ions of the HTLNQIDSVK peptide. Which are usually the ones that can be used to make an identification.

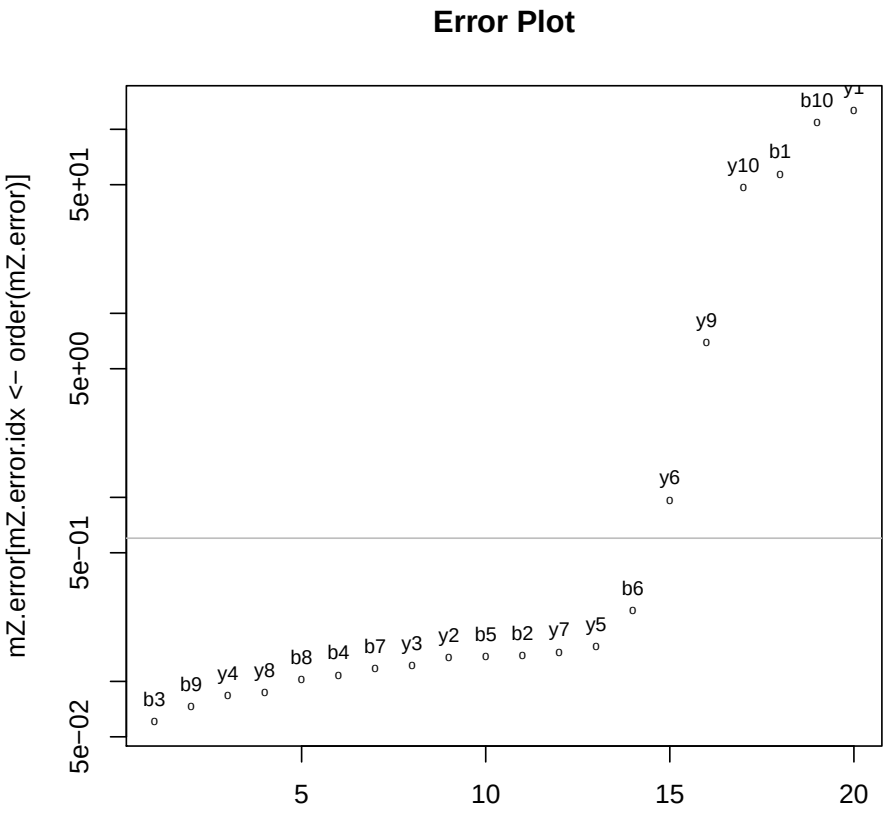
```
> Hydrogen<-1.007825
> (fi.HTLNQIDSVK.1<-fragmentIon('HTLNQIDSVK'))[[1]]

           b         y         c         z
1   138.0662  147.1128  155.0927  130.0863
2   239.1139  246.1812  256.1404  229.1547
3   352.1979  333.2132  369.2245  316.1867
4   466.2409  448.2402  483.2674  431.2136
5   594.2994  561.3242  611.3260  544.2977
6   707.3835  689.3828  724.4100  672.3563
7   822.4104  803.4258  839.4370  786.3992
8   909.4425  916.5098  926.4690  899.4833
9  1008.5109 1017.5575 1025.5374 1000.5309
10 1136.6058 1154.6164 1153.6324 1137.5899

> (fi.HTLNQIDSVK.2<-(fi.HTLNQIDSVK.1[[1]] + Hydrogen) / 2)

           b         y         c         z
1   69.53701  74.06031  78.05028  65.54704
2  120.06085 123.59452 128.57412 115.08124
3  176.60288 167.11053 185.11615 158.59726
4  233.62434 224.62400 242.13761 216.11073
5  297.65363 281.16603 306.16691 272.65276
6  354.19566 345.19532 362.70894 336.68205
7  411.70913 402.21679 420.22241 393.70351
8  455.22515 458.75882 463.73842 450.24554
9  504.75935 509.28266 513.27262 500.76938
10 568.80683 577.81211 577.32010 569.29884
```

4.3 Peptide Sequence – Fragment Ion Matching

Given a peptide sequence and a tandem mass spectrum. For the assignment of a canditate peptide an in-silico fragment ion spectra `fi` is computed. The function `findNN` determines for each fragment ion the closesed peak in the MS2. If the difference between the in-silico mass and the measured mass is inside the 'accuracy' mass window of the mass spec device the in-silico fragment ion is considered as potential hit.

```
>     peptideSequence<-'HTLNQIDSVK'
>     spec<-list(scans=1138,
+         title="178: (rt=22.3807) [20080816_23_fetuin_160.RAW]",
+         rtinseconds=1342.8402,
+         charge=2,
+         mZ=c(195.139940, 221.211970, 239.251780, 290.221750,
+     316.300770, 333.300050, 352.258420, 448.384360, 466.348830,
+     496.207570, 509.565910, 538.458310, 547.253380, 556.173940,
```

```
+     560.358050, 569.122080, 594.435500, 689.536940, 707.624790,
+     803.509240, 804.528220, 822.528020, 891.631250, 909.544400,
+     916.631600, 973.702160, 990.594520, 999.430580, 1008.583600,
+     1017.692500, 1027.605900),
+         intensity=c(931.8, 322.5, 5045, 733.9, 588.8, 9186, 604.6,
+     1593, 531.8, 520.4, 976.4, 410.5, 2756, 2279, 5819, 2.679e+05,
+     1267, 1542, 979.2, 9577, 3283, 9441, 1520, 1310, 1.8e+04,
+     587.5, 2685, 671.7, 3734, 8266, 3309))
>     fi<-fragmentIon(peptideSequence)
>     n<-nchar(peptideSequence)
>     by.mZ<-c(fi[[1]]$b, fi[[1]]$y)
>     by.label<-c(paste("b",1:n,sep=''), paste("y",n:1,sep=''))
>     # should be a R-core function as findInterval!
>     idx<-findNN(by.mZ, spec$mZ)
>     mZ.error<-abs(spec$mZ[idx]-by.mZ)
>     plot(mZ.error[mZ.error.idx<-order(mZ.error)],
+         main="Error Plot",
+         pch='o',
+         cex=0.5,
+         sub='The error cut-off is 0.6Da (grey line).',
+         log='y')
>     abline(h=0.6,col='grey')
>     text(1:length(by.label),
+         mZ.error[mZ.error.idx],
+         by.label[mZ.error.idx],
+         cex=0.75,pos=3)
```



The error cut−off is 0.6Da (grey line).

The graphic above is showing the mass error of the assingment between the MS2 spec and the singly charged fragment ions of HTLNQIDSVK. The function psm is doing the peptide sequence matching. Of course, the more theoretical ions match (up to a small error tolerance, given by the system) the actually measured ion series, the more likely it is, that the measured spectrum indeed is from the inferred peptide (and therefore the protein is identified)

4.4 Modifications

```
> library(protViz)
> ptm.0<-cbind(AA="-",
+     mono=0.0, avg=0.0, desc="unmodified", unimodAccID=NA)
> ptm.616<-cbind(AA='S',
+     mono=-27.010899, avg=NA, desc="Substituition",
+     unimodAccID=616)
> ptm.651<-cbind(AA='N',
+     mono=27.010899, avg=NA, desc="Substituition",
+     unimodAccID=651)
> m<-as.data.frame(rbind(ptm.0, ptm.616, ptm.651))
> genMod(c('TAFDEAIAELDTLNEESYK','TAFDEAIAELDTLSEESYK'), m$AA)

[[1]]
[1] "000000000000000000" "000000000000200000" "000000000000000100"
[4] "000000000000100200"
```

```
[[2]]
[1] "0000000000000000000" "0000000000000100000" "0000000000000000100"
[4] "0000000000000100100"

> fi<-fragmentIon(c('TAFDEAIAELDTLSEESYK',
+     'TAFDEAIAELDTLNEESYK', 'TAFDEAIAELDTLSEESYK',
+     'TAFDEAIAELDTLNEESYK'),
+         modified=c('0000000000000200000',
+         '0000000000000100000', '0000000000000000000',
+         '0000000000000000000'),
+     modification=m$mono)
>
> #bh<-c('TAFDEAIAELDTLNEESYK', 'TAFDEAIAELDTLSEESYK')
> #fi<-fragmentIon(rep('HTLNQIDSVK',2),
> #    modified=c('0000000100','0000000000'),
> #    modification=m[,2])
```

4.5 Labeling Peaklists

The labeling of the spectra can be done with the peakplot function.

```
> data(msms)
> op<-par(mfrow=c(2,1))
> peakplot("TAFDEAIAELDTLNEESYK", msms[[1]])

$mZ.Da.error
 [1] 232.331344 161.294234  14.225824  -0.032616  -0.143306   0.032244
 [7]   0.054604  -0.004076  -0.071746  -0.084536  -0.097076  -0.038856
[13]  -0.061816   0.004554  -0.122336  -0.139626  -1.071256         NA
[19]         NA 187.273499  24.210169   0.048669   0.177779   0.027939
[25]   0.049579   0.052379   0.044579   0.036749   0.043189  -0.035101
[31]  -0.061011   0.000729  -0.092081   2.011029  -8.412111   7.195579
[37]         NA         NA 215.304795 144.267685  -2.800725 -17.059165
[43]   2.034875   2.264105   4.008125   1.292875  -0.003965 -13.612585
[49]  -0.060925 -17.065405   3.897535   3.000405 -17.148885 -17.166175
[55] -18.097805         NA         NA 204.300048  41.236718  17.075218
[61]  -0.843372  -1.091812   0.129908  17.078928  -0.372162 -16.539502
[67]  -1.044962  -1.000952  -1.409062  -2.995122  16.934468  19.037578
[73]   8.614438  24.222128         NA         NA

$mZ.ppm.error
 [1]  2.276532e+06  9.318407e+05  4.443342e+04 -7.494702e+01 -2.539851e+02
 [6]  5.075660e+01  7.296574e+01 -4.974443e+00 -7.564705e+01 -7.963713e+01
[11] -8.250960e+01 -3.041352e+01 -4.445040e+01  3.026484e+00 -7.488007e+01
[16] -7.920687e+01 -5.791093e+02            NA            NA  1.272993e+06
[21]  7.805297e+04  1.225277e+02  3.378218e+02  4.263587e+01  6.444386e+01
[26]  5.935833e+01  4.532837e+01  3.345395e+01  3.564687e+01 -2.618263e+01
```

```
[31] -4.321937e+01  4.781134e-01 -5.770282e+01  1.165934e+03 -4.572174e+03
[36]  3.621478e+03            NA            NA  1.808046e+06  7.588299e+05
[41] -8.306147e+03 -3.772366e+04  3.500821e+03  3.470990e+03  5.236793e+03
[46]  1.545734e+03 -4.106862e+00 -1.262129e+04 -5.104441e+01 -1.318183e+04
[51]  2.768725e+03  1.971690e+03 -1.038832e+04 -9.644849e+03 -9.694247e+03
[56]            NA            NA  1.570497e+06  1.406678e+05  4.491332e+04
[61] -1.656190e+03 -1.710589e+03  1.726789e+02  1.973544e+04 -3.850849e+02
[66] -1.529356e+04 -8.747728e+02 -7.562373e+02 -1.010347e+03 -1.986529e+03
[71]  1.072648e+04  1.114745e+04  4.725878e+03  1.229618e+04            NA
[76]            NA

$idx
 [1]   1   1   1   3  14  21  38  49  64  87  91  97 102 106 110 113 115 117 117
[20]   1   1   2  12  25  41  53  70  89  94  99 104 107 108 111 114 116 117 117
[39]   1   1   1   3  16  24  41  52  67  88  93  97 104 107 110 113 115 117 117
[58]   1   1   2  11  22  40  53  68  88  93  98 103 106 108 111 114 116 117 117

$label
 [1] "b1"  "b2"  "b3"  "b4"  "b5"  "b6"  "b7"  "b8"  "b9"  "b10" "b11" "b12"
[13] "b13" "b14" "b15" "b16" "b17" "b18" "b19" "y1"  "y2"  "y3"  "y4"  "y5"
[25] "y6"  "y7"  "y8"  "y9"  "y10" "y11" "y12" "y13" "y14" "y15" "y16" "y17"
[37] "y18" "y19" "c1"  "c2"  "c3"  "c4"  "c5"  "c6"  "c7"  "c8"  "c9"  "c10"
[49] "c11" "c12" "c13" "c14" "c15" "c16" "c17" "c18" "c19" "z1"  "z2"  "z3"
[61] "z4"  "z5"  "z6"  "z7"  "z8"  "z9"  "z10" "z11" "z12" "z13" "z14" "z15"
[73] "z16" "z17" "z18" "z19"

$score
[1] -1

$sequence
[1] "TAFDEAIAELDTLNEESYK"

$fragmentIon
           b         y         c         z
1   102.0550  147.1128  119.0815  130.0863
2   173.0921  310.1761  190.1186  293.1496
3   320.1605  397.2082  337.1870  380.1816
4   435.1874  526.2508  452.2140  509.2242
5   564.2300  655.2933  581.2566  638.2668
6   635.2671  769.3363  652.2937  752.3097
7   748.3512  882.4203  765.3777  865.3938
8   819.3883  983.4680  836.4148  966.4415
9   948.4309 1098.4950  965.4574 1081.4684
10 1061.5149 1211.5790 1078.5415 1194.5525
11 1176.5419 1340.6216 1193.5684 1323.5951
12 1277.5896 1411.6587 1294.6161 1394.6322
13 1390.6736 1524.7428 1407.7002 1507.7162
```

```
14 1504.7165 1595.7799 1521.7431 1578.7533
15 1633.7591 1724.8225 1650.7857 1707.7959
16 1762.8017 1839.8494 1779.8283 1822.8229
17 1849.8338 1986.9178 1866.8603 1969.8913
18 2012.8971 2057.9549 2029.9236 2040.9284
19 2140.9920 2159.0026 2158.0186 2141.9761
```

```
> peakplot("TAFDEAIAELDTLSEESYK", msms[[2]])
```

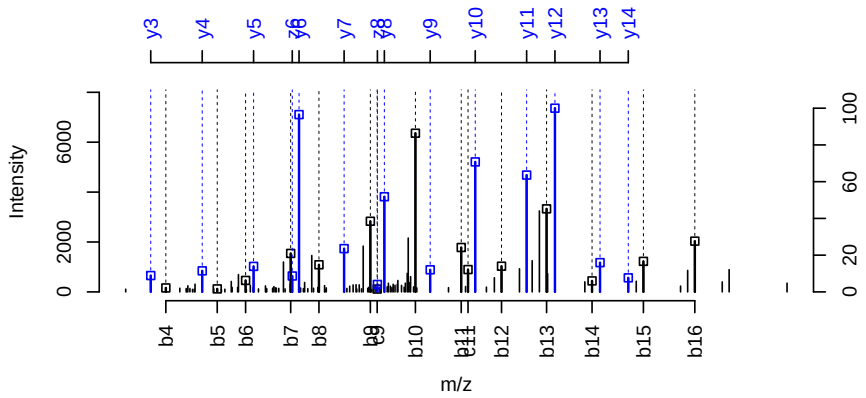
```
$mZ.Da.error
 [1] 245.264254 174.227144  27.158734  14.444434   0.021404  -0.111266
 [7]  -0.039926  -0.021626  -0.121916  -8.079236  -0.158376  -0.153156
[13]  -0.094316  -0.022946  -0.186736  -0.092226  -0.120456         NA
[19]         NA 200.206409  37.143079   0.078909   0.062269   0.129769
[25]   0.103729   0.060869  -0.051451 -18.048351  -0.027511  -0.025601
[31]  -0.006211   0.020529  -0.048781  -0.024771  -9.166311   6.953579
[37]         NA         NA 228.237705 157.200595  10.132185  -2.582115
[43]   1.626855   2.722405   9.009025  -1.130895   1.216385  13.347315
[49]  -3.671525   0.960295 -17.120865   3.020205 -17.213285 -17.118775
[55] -17.147005         NA         NA 217.232958  54.169628  17.105458
[61]  -0.833452  -1.260332  -0.899352  -3.098942  -1.173512  -1.021802
[67]  -0.939162  -1.007752  -1.377062  -3.022622  16.977768  17.001778
[73]   7.860238  23.980128         NA         NA
```

```
$mZ.ppm.error
 [1]  2.403257e+06  1.006558e+06  8.482850e+04  3.319130e+04  3.793488e+01
 [6] -1.751484e+02 -5.335196e+01 -2.639286e+01 -1.285450e+02 -7.611043e+03
[11] -1.346114e+02 -1.198789e+02 -6.782037e+01 -1.552813e+01 -1.162198e+02
[16] -5.313198e+01 -6.608212e+01            NA            NA  1.360904e+06
[21]  1.197483e+05  1.986591e+02  1.183257e+02  1.980319e+02  1.397352e+02
[26]  7.115774e+01 -5.379332e+01 -1.684426e+04 -2.322450e+01 -1.948903e+01
[31] -4.485617e+00  1.370673e+01 -3.109508e+01 -1.458996e+01 -5.056331e+03
[36]  3.547913e+03            NA            NA  1.916651e+06  8.268554e+05
[41]  3.004915e+04 -5.709941e+03  2.798859e+03  4.173588e+03  1.177069e+04
[46] -1.352074e+03  1.259905e+03  1.237534e+04 -3.076091e+03  7.417604e+02
[51] -1.216230e+04  2.020566e+03 -1.060078e+04 -9.766434e+03 -9.319787e+03
[56]            NA            NA  1.669915e+06  1.847849e+05  4.499286e+04
[61] -1.636709e+03 -1.974616e+03 -1.239974e+03 -3.696333e+03 -1.249174e+03
[66] -9.690310e+02 -8.043928e+02 -7.772361e+02 -1.006903e+03 -2.041339e+03
[71]  1.094110e+04  1.011538e+04  4.376983e+03  1.234257e+04            NA
[76]            NA
```

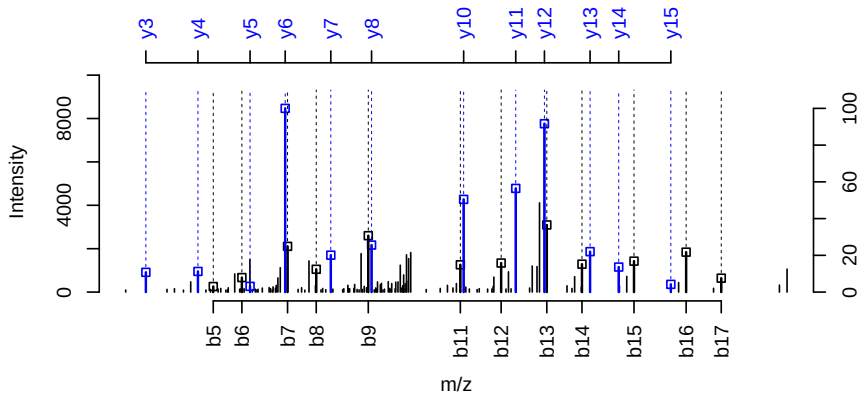
```
$idx
 [1]   1   1   1   3  11  20  39  45  64  90  96 106 116 121 126 129 131 134 134
[20]   1   1   2   7  24  38  49  65  90  97 110 115 122 123 127 130 132 134 134
[39]   1   1   1   3  13  23  40  47  67  91  98 108 116 122 126 129 131 134 134
[58]   1   1   2   6  21  36  47  62  90  95 108 113 121 123 127 130 132 134 134
```

```
$label
 [1] "b1"  "b2"  "b3"  "b4"  "b5"  "b6"  "b7"  "b8"  "b9"  "b10" "b11" "b12"
[13] "b13" "b14" "b15" "b16" "b17" "b18" "b19" "y1"  "y2"  "y3"  "y4"  "y5"
[25] "y6"  "y7"  "y8"  "y9"  "y10" "y11" "y12" "y13" "y14" "y15" "y16" "y17"
[37] "y18" "y19" "c1"  "c2"  "c3"  "c4"  "c5"  "c6"  "c7"  "c8"  "c9"  "c10"
[49] "c11" "c12" "c13" "c14" "c15" "c16" "c17" "c18" "c19" "z1"  "z2"  "z3"
[61] "z4"  "z5"  "z6"  "z7"  "z8"  "z9"  "z10" "z11" "z12" "z13" "z14" "z15"
[73] "z16" "z17" "z18" "z19"

$score
[1] -1

$sequence
[1] "TAFDEAIAELDTLSEESYK"

$fragmentIon
           b         y         c         z
1   102.0550  147.1128  119.0815  130.0863
2   173.0921  310.1761  190.1186  293.1496
3   320.1605  397.2082  337.1870  380.1816
4   435.1874  526.2508  452.2140  509.2242
5   564.2300  655.2933  581.2566  638.2668
6   635.2671  742.3254  652.2937  725.2988
7   748.3512  855.4094  765.3777  838.3829
8   819.3883  956.4571  836.4148  939.4306
9   948.4309 1071.4841  965.4574 1054.4575
10 1061.5149 1184.5681 1078.5415 1167.5416
11 1176.5419 1313.6107 1193.5684 1296.5842
12 1277.5896 1384.6478 1294.6161 1367.6213
13 1390.6736 1497.7319 1407.7002 1480.7053
14 1477.7056 1568.7690 1494.7322 1551.7424
15 1606.7482 1697.8116 1623.7748 1680.7850
16 1735.7908 1812.8385 1752.8174 1795.8120
17 1822.8229 1959.9069 1839.8494 1942.8804
18 1985.8862 2030.9440 2002.9127 2013.9175
19 2113.9811 2131.9917 2131.0077 2114.9652
```

> par(op)



YK / 1799: Scan 3246 (rt=67.4676) [/p474/Proteomics/ORBI_1/jonas_20080530_bhdaten_doro/20071028_b



YK / 1842: Scan 3353 (rt=70.3158) [/p474/Proteomics/ORBI_1/jonas_20080530_bhdaten_doro/20071028_b

5 Quantification

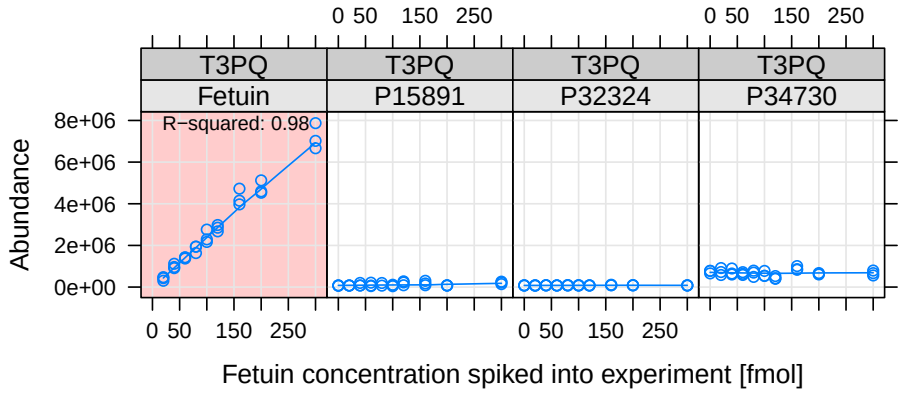
For an overview on Quantitative Proteomics read [1, 2]. The authors are aware that meaningful statistics usually require much higher number of biological replicates. In almost all cases there are not more than three to six repitions. For the moment there are limited options due to the availabilty of machine time and the limits of the technologies.

5.1 Label-free methods on protein level

The data set `fetuinLFQ` contains a subset of our results descriped in [3]. The example below shows a visualization using trellis plots. It graphs the abundance of four protein in dependency from the fetuin concentration spiked into the sample.

```
> library(lattice)
> data(fetuinLFQ)
> cv<-1-1:7/10
> t<-trellis.par.get("strip.background")
> t$col<-(rgb(cv,cv,cv))
> trellis.par.set("strip.background",t)
> print(xyplot(abundance~conc|prot*method,
+     groups=prot,
```

```
+     xlab="Fetuin concentration spiked into experiment [fmol]",
+     ylab="Abundance",
+     aspect=1,
+     data=fetuinLFQ$t3pq[fetuinLFQ$t3pq$prot
+         %in% c('Fetuin', 'P15891', 'P32324', 'P34730'),],
+     panel = function(x, y, subscripts, groups) {
+         if (groups[subscripts][1] == "Fetuin")  {
+             panel.fill(col="#ffcccc")
+         }
+         panel.grid(h=-1,v=-1)
+         panel.xyplot(x, y)
+         panel.loess(x,y, span=1)
+         if (groups[subscripts][1] == "Fetuin")  {
+             panel.text(min(fetuinLFQ$t3pq$conc),
+                 max(fetuinLFQ$t3pq$abundance),
+                 paste("R-squared:",
+                 round(summary(lm(x~y))$r.squared,2)),
+                 cex=0.75,
+                 pos=4)
+         }
+     }
+ ))
```

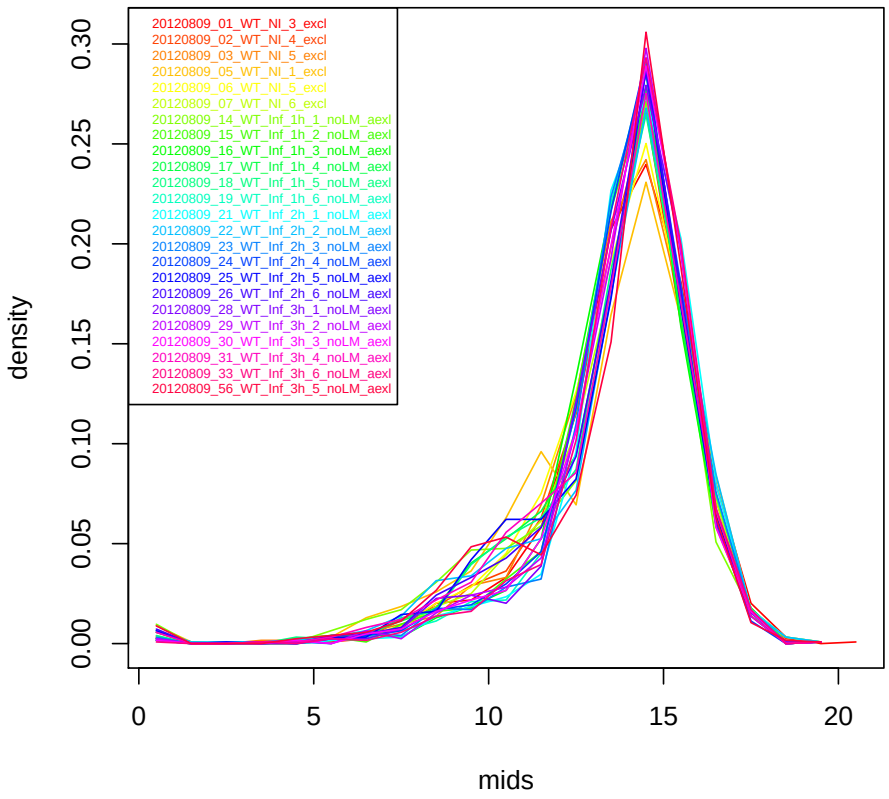


The plot shows the estimated concentration of the four proteins using the top three most intense peptides. The Fetuin peptides are spiked in with increasing concentration while the three other yeast proteins are kept stable in the background.

5.2 pgLFQ – LCMS based label-free quantification

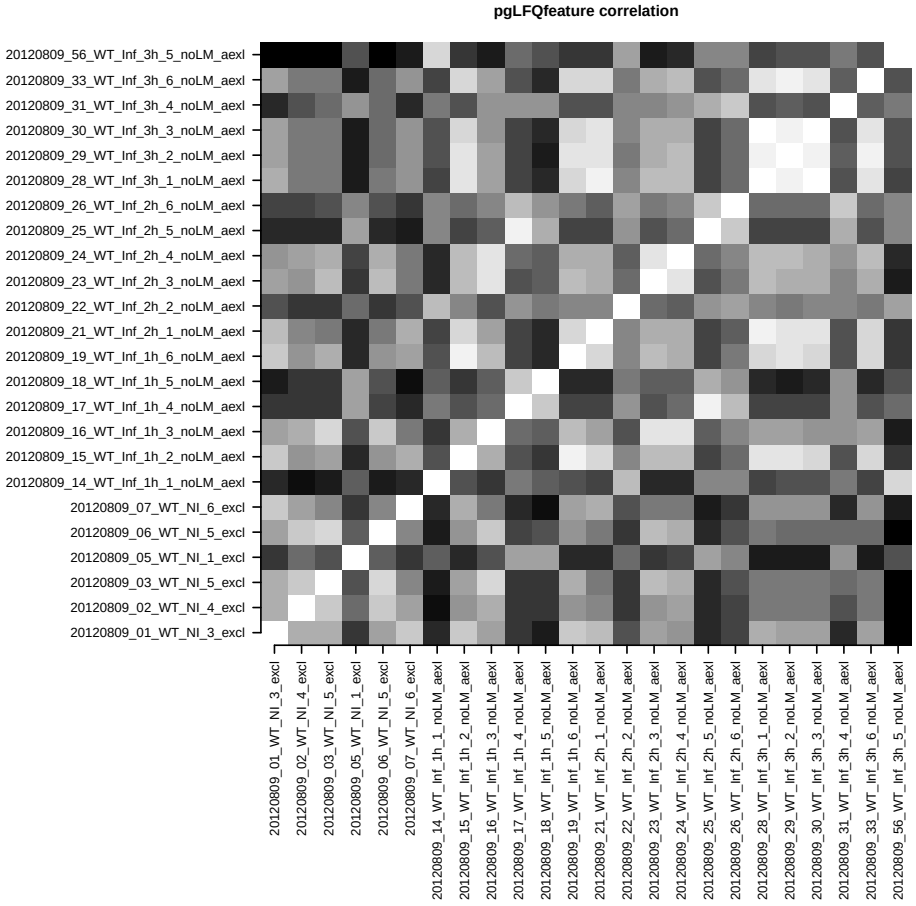
LCMS based label-free quantification is a very popular method to extract relative quantitative information from mass spectrometry experiments. At the FGCZ we use the software ProgenesisLCMS for this workflow `http://www.nonlinear.com/products/progenesis/lc-ms/overview/`. Progenesis is a graphical software which does the aligning between several LCMS experiments, extracts signal intensities from LCMS maps and annotates the mastermap with peptide and protein labels.

```
> data(pgLFQfeature)
> data(pgLFQprot)
> featureDensityPlot<-function(data, n=ncol(data), nbins=30){
+     my.col<-rainbow(n);
+     mids<-numeric()
+     density<-numeric()
+     for (i in 1:n) {
+         h<-hist(data[,i],nbins, plot=FALSE)
+         mids<-c(mids, h$mids)
+         density<-c(density, h$density)
+     }
+     plot(mids,density, type='n')
+     for (i in 1:n) {
+         h<-hist(data[,i],nbins, plot=FALSE)
+         lines(h$mids,h$density, col=my.col[i])
+     }
+     legend("topleft", names(data), cex=0.5,
+         text.col=my.col
+     )
+ }
> par(mfrow=c(1,1));
> featureDensityPlot(asinh(pgLFQfeature$"Normalized abundance"),
+     nbins=25)
```



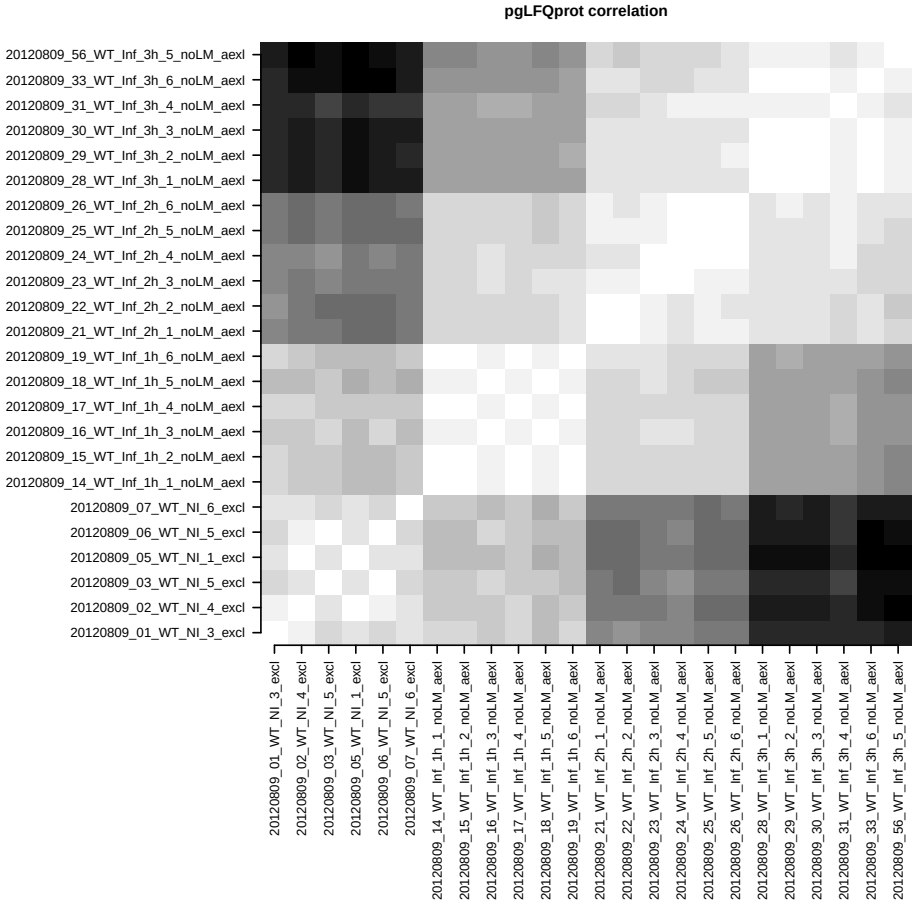
The `featureDensityPlot` shows the normalized signal intensity distribution (asinh transformed) over 24 LCMS runs which are aligned in this experiment.

```
> op<-par(mfrow=c(1,1),mar=c(18,18,4,1),cex=0.5)
> samples<-names(pgLFQfeature$"Normalized abundance")
> image(cor(asinh(pgLFQfeature$"Normalized abundance")),
+     col=gray(seq(0,1,length=20)),
+     main='pgLFQfeature correlation',
+     axes=FALSE)
> axis(1,at=seq(from=0, to=1,
+     length.out=length(samples)),
+     labels=samples, las=2)
> axis(2,at=seq(from=0, to=1,
+     length.out=length(samples)), labels=samples, las=2)
> par(op)
```



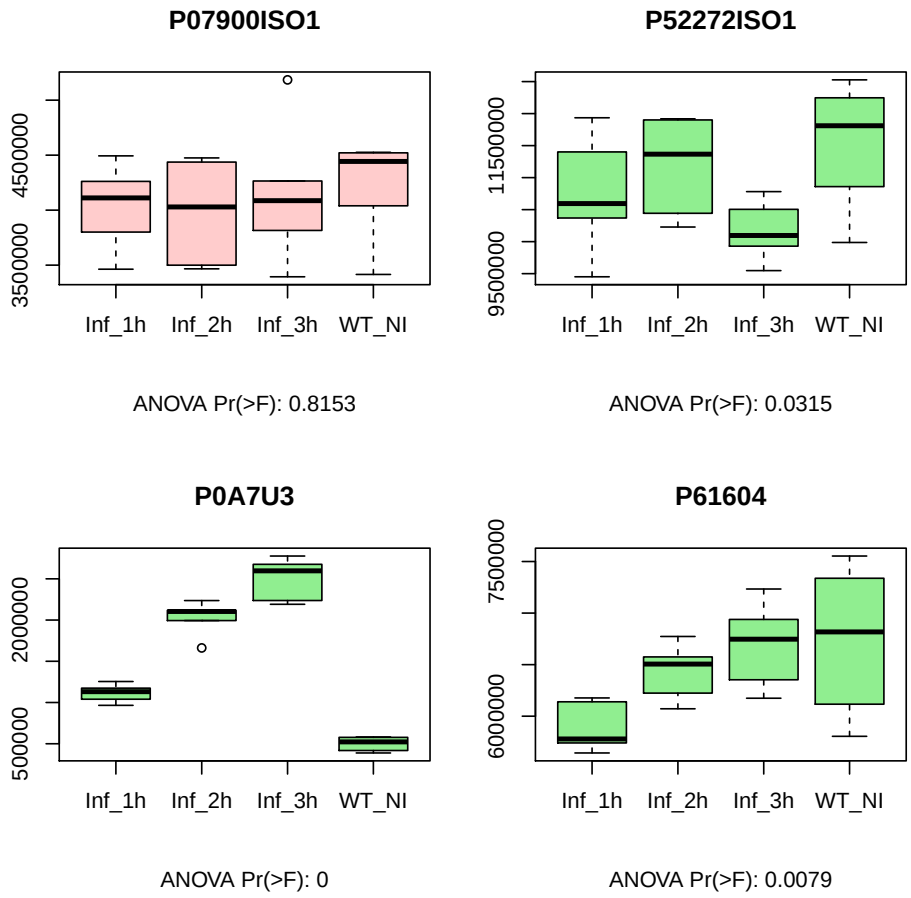
This image plot shows the correlation between runs on feature level (values are asinh transformed). White is perfect correlation while black indicates a poor correlation.

```
> op<-par(mfrow=c(1,1),mar=c(18,18,4,1),cex=0.5)
> image(cor(asinh(pgLFQprot$"Normalized abundance")),
+     main='pgLFQprot correlation',
+     axes=FALSE,
+     col=gray(seq(0,1,length=20)))
> axis(1,at=seq(from=0, to=1,
+     length.out=length(samples)), labels=samples, las=2)
> axis(2,at=seq(from=0, to=1,
+     length.out=length(samples)), labels=samples, las=2)
> par(op)
```



This figure shows the correlation between runs on protein level (values are asinh transformed). White is perfect correlation while black indicates a poor correlation. Stricking is the fact that the six biological replicates for each condition cluster very well.

```
> par(mfrow=c(2,2),mar=c(6,3,4,1))
> ANOVA<-pgLFQaov(pgLFQprot$"Normalized abundance",
+      groups=as.factor(pgLFQprot$grouping),
+      names=pgLFQprot$output$Accession,
+      idx=c(15,16,196,107),
+      plot=TRUE)
```



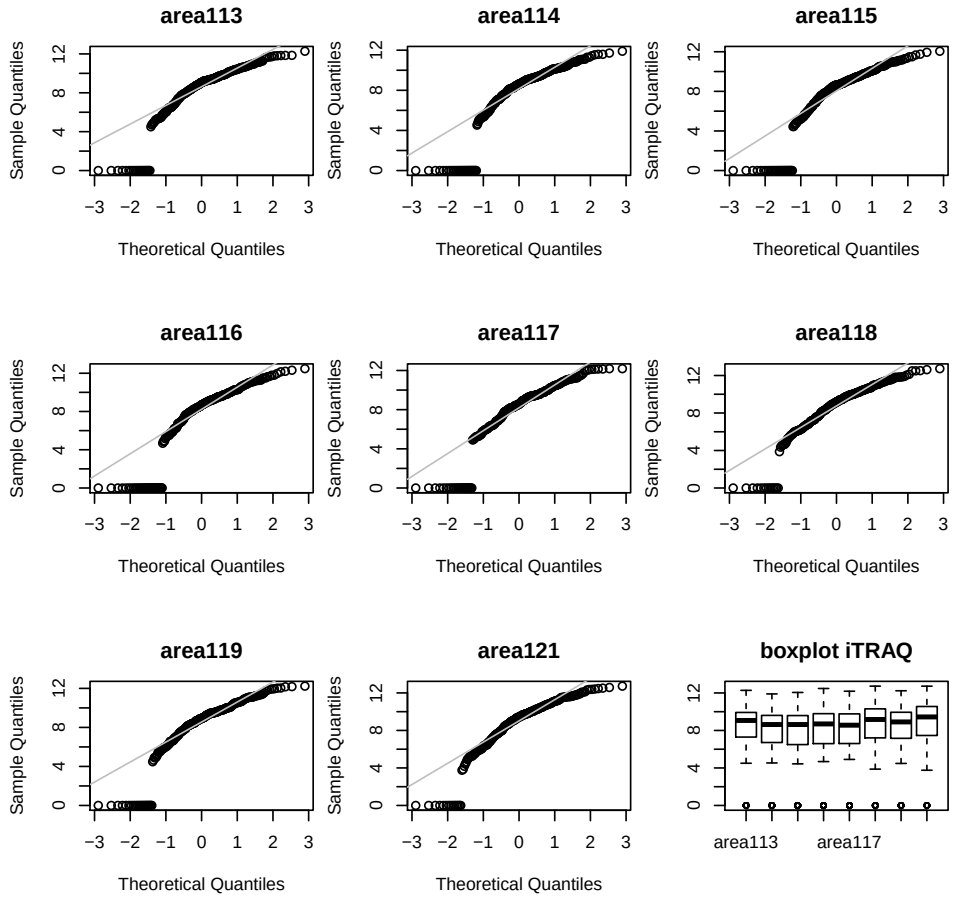
This figure shows the result for four proteins which either differ significantly in expression accross conditions (green boxplots) using an analysis of variance test, or non differing protein expression (red boxplot).

5.3 iTRAQ – Two Group Analysis

The data for the next section is an iTRAQ-8-plex experiment where two conditions are compared (each condition has 4 biological replicates)

5.3.1 Sanity Check

```
> data(iTRAQ)
> x<-rnorm(100)
> par(mfrow=c(3,3),mar=c(6,4,3,0.5));
> for (i in 3:10){
+     qqnorm(asinh(iTRAQ[,i]),
+         main=names(iTRAQ)[i])
+     qqline(asinh(iTRAQ[,i]), col='grey')
+ }
> b<-boxplot(asinh(iTRAQ[,c(3:10)]), main='boxplot iTRAQ')
```

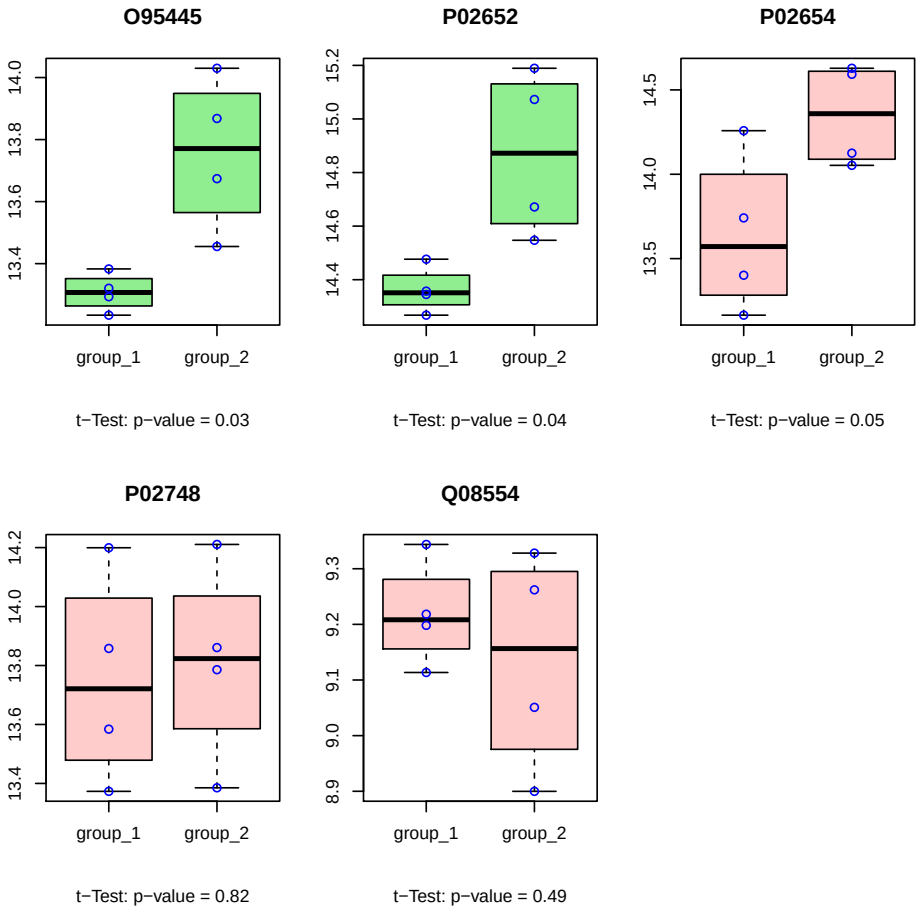


A first quality check to see if all reporter ion channels are having the same distributions. Shown in the figure are Q-Q plots of the individual reporter channels against a normal distribution. The last is a boxplot for all individual channels.

5.3.2 On Protein Level

```
> data(iTRAQ)
> group1Protein<-numeric()
> group2Protein<-numeric()
> for (i in c(3,4,5,6))
+     group1Protein<-cbind(group1Protein,
+         asinh(tapply(iTRAQ[,i], paste(iTRAQ$prot), sum, na.rm=TRUE)))
> for (i in 7:10)
+     group2Protein<-cbind(group2Protein,
+         asinh(tapply(iTRAQ[,i], paste(iTRAQ$prot), sum, na.rm=TRUE)))
> par(mfrow=c(2,3),mar=c(6,3,4,1))
> for (i in 1:nrow(group1Protein)){
+     boxplot.color="#ffcccc"
+     tt.p_value<-t.test(as.numeric(group1Protein[i,]),
+         as.numeric(group2Protein[i,]))$p.value
+
+     if (tt.p_value < 0.05)
```

```
+         boxplot.color='lightgreen'
+
+     b<-boxplot(as.numeric(group1Protein[i,]),
+         as.numeric(group2Protein[i,]),
+         main=row.names(group1Protein)[i],
+         sub=paste("t-Test: p-value =", round(tt.p_value,2)),
+         col=boxplot.color,
+         axes=FALSE)
+     axis(1, 1:2, c('group_1','group_2')); axis(2); box()
+
+     points(rep(1,b$n[1]), as.numeric(group1Protein[i,]), col='blue')
+     points(rep(2,b$n[2]), as.numeric(group2Protein[i,]), col='blue')
+ }
```



This figure shows five proteins which are tested if they differ accross conditions using the four biological replicates with a t-test.

5.3.3 On Peptide Level

The same can be done on peptide level using the `protViz` function `iTRAQ2GroupAnalysis`.

```
> data(iTRAQ)
> q<-iTRAQ2GroupAnalysis(data=iTRAQ,
```

```
+     group1=c(3,4,5,6),
+     group2=7:10,
+     INDEX=paste(iTRAQ$prot,iTRAQ$peptide),
+     plot=FALSE)
> q[1:10,]
```

	name	p_value	Group1.area113	Group1.area114	Group1.area115	Group1.area116	Group2.area117	Group2.area118	Group2.area119	Group2.area121
1	O95445 AFLLTPR	0.056	1705.43	1459.10	770.65	3636.40	3063.48	4046.73	2924.49	5767.87
2	O95445 DGLCVPR	0.161	2730.41	1852.90	1467.65	2266.88	2269.57	3572.32	2064.82	2208.92
3	O95445 MKDGLCVPR	0.039	28726.38	15409.81	19050.13	58185.02	51416.05	70721.05	38976.42	60359.72
4	O95445 NQEACELSNN	0.277	4221.31	4444.28	2559.23	6859.71	5545.12	11925.66	6371.50	15656.92
5	O95445 SLTSCLDSK	0.036	20209.66	14979.02	12164.94	37572.56	30687.57	39176.99	34417.66	54439.22
6	P02652 AGTELVNFLSYFVELGTQPA	0.640	4504.97	4871.88	2760.53	9213.41	6728.62	14761.96	7796.29	18681.60
7	P02652 AGTELVNFLSYFVELGTQPAT	0.941	67308.30	46518.21	33027.14	111629.30	94531.76	168775.00	83526.72	168032.50
8	P02652 AGTELVNFLSYFVELGTQPATQ	0.338	4661.54	3971.82	2564.39	8269.73	6045.30	13724.92	7426.84	17214.87
9	P02652 EPCVESLVSQYFQTVTDYGK	0.115	4544.56	4356.51	2950.48	6357.90	6819.99	10265.84	7012.92	14279.22
10	P02652 EQLTPLIK	0.053	24596.42	22015.94	18424.56	49811.91	33197.47	67213.62	40030.86	87343.38

6 Pressure Profiles QC

A common problem with mass spec setup is the pure reliability of the high pressure pump. The following graphics provide visualizations for quality control.

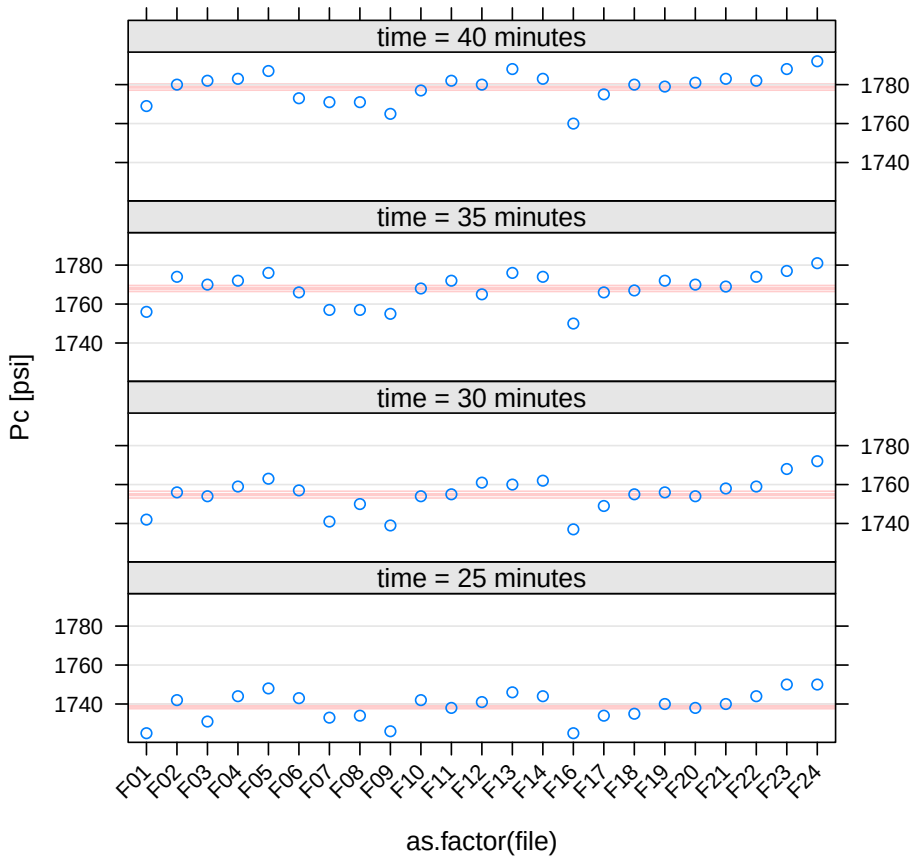
On overview of the pressure profile data can be seen by using the ppp function.

```
> data(pressureProfile)
> ppp(pressureProfile)
```

The lines plots the pressure profiles data on a scatter plot 'Pc' versus 'time' grouped by time range (no figure because of too many data items).

The Trellis `xyplot` shows the Pc development over each instrument run to a specified relative run time (25,30,...).

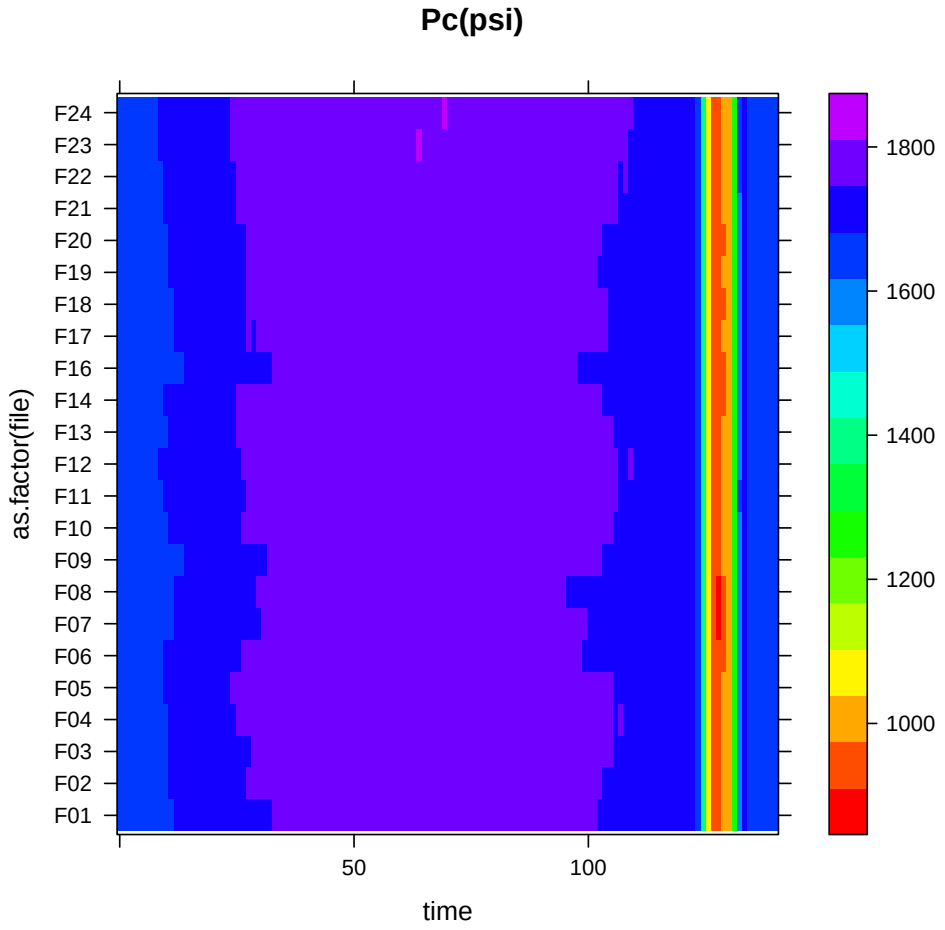
```
> pp.data<-pps(pressureProfile, time=seq(25,40,by=5))
> print(xyplot(Pc ~ as.factor(file) | paste("time =",
+     as.character(time), "minutes"),
+     panel = function(x, y){
+         m<-sum(y)/length(y)
+         m5<-(max(y)-min(y))*0.05
+         panel.abline(h=c(m-m5,m,m+m5),
+             col=rep("#ffcccc",3),lwd=c(1,2,1))
+         panel.grid(h=-1, v=0)
+         panel.xyplot(x, y)
+     },
+     ylab='Pc [psi]',
+     layout=c(1,4),
+     sub='The three red lines indicate the average plus min 5%.',
+     scales = list(x = list(rot = 45)),
+     data=pp.data))
```



The three red lines indicate the average plus min 5%.

While each panel in the `xyplot` above shows the data to a given point in time, we try to use the levelplot to get an overview of the whole pressure profile data.

```
> pp.data<-pps(pressureProfile, time=seq(0,140,length=128))
> print(levelplot(Pc ~ time * as.factor(file),
+     main='Pc(psi)',
+     data=pp.data,
+     col.regions=rainbow(100)[1:80]))
```



References

[1] M.˜Bantscheff, S.˜Lemeer, M.˜M. Savitski, and B.˜Kuster. Quantitative mass spectrometry in proteomics: critical review update from 2007 to the present. *Anal Bioanal Chem*, 404(4):939–965, Sep 2012.

[2] S.˜Cappadona, P.˜R. Baker, P.˜R. Cutillas, A.˜J. Heck, and B.˜van Breukelen. Current challenges in software solutions for mass spectrometry-based quantitative proteomics. *Amino Acids*, 43(3):1087–1108, Sep 2012.

[3] J.˜Grossmann, B.˜Roschitzki, C.˜Panse, C.˜Fortes, S.˜Barkow-Oesterreicher, D.˜Rutishauser, and R.˜Schlapbach. Implementation and evaluation of relative and absolute quantification in shotgun proteomics with label-free methods. *J Proteomics*, 73(9):1740–1746, Aug 2010. [DOI:10.1016/j.jprot.2010.05.011] [PubMed:20576481].

[4] P.˜Roepstorff and J.˜Fohlman. Proposal for a common nomenclature for sequence ions in mass spectra of peptides. *Biomed. Mass Spectrom.*, 11(11):601, Nov 1984. [DOI:10.1002/bms.1200111109] [PubMed:6525415].