

A computational pipeline to generate and analyze proBAM files

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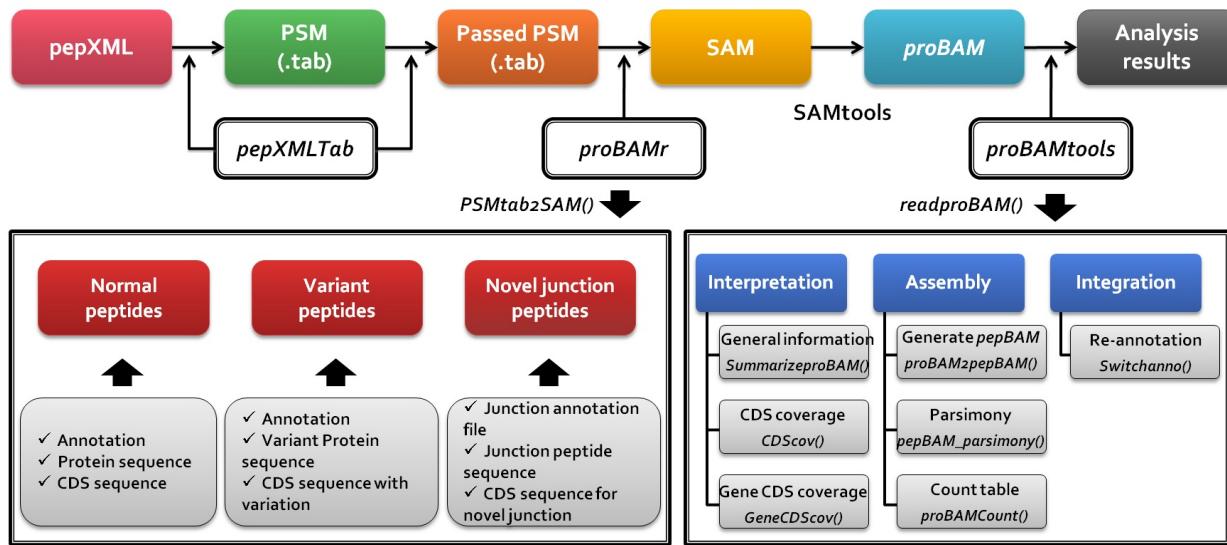
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1 Introduction

Recent advances of sequencing technologies have reformed our conception of genomic data analysis, storage and interpretation, instigating more research interest in exploring human proteome at a parallel scale. Shotgun proteomics holds this promise by surveying proteome both qualitatively and quantitatively. Over the last years large amount of proteomics data has been accumulated, an emerging demand is to combine these efforts to catalogue the wide dynamic range of protein expression and complexity of alternative isoforms. However, this task is daunting due to the fact that different studies use varying databases, search engines and assembly tools. Such a challenge calls for an efficient approach of integrating data from different proteomics studies and even with genomic data.

Here we provide a computational pipeline, that maps identified PSMs to the genome in BAM format, a binary format for efficient data storage and fast access in genomic research field. This method differs from other approaches because of its ability of building connections between peptide and genomic location and simultaneously maintaining spectra count information. PSMs are aligned under the same coordination framework regardless of the annotation systems (e.g. RefSeq, ENSEMBL) of the input proteomics data, which enables flexible protein assembly switch between different annotation or at different level (gene or protein). When genomic/transcriptomic information of the same individual is available, this approach allows the co-analysis with -omics data together. In Figure 1, we illustrated the pipeline for generating and analyzing proBAM files. This document provides a step by step tutorial using examples.

Figure 1: A computational pipeline to generate and analyze proBAM files



2 Building proBAM files using R package *probAMr*

2.1 Preparing annotation files

To map proteomics data to the genome, numerous pieces of genome annotation information are needed, such as genome elements region boundary, protein coding sequence and protein sequence et al. It is possible to manually download these data from different public resources (e.g. NCBI, UCSC and ENSEMBL) and then parse them to an appropriate format. To make this process more efficient and autonomous, we provide functions to prepare the gene/transcript annotation files from UCSC, ENSEMBL and GENCODE. The `PrepareAnnotationRefseq` and `PrepareAnnotationEnsembl` were included in another R package `customProDB` <http://bioconductor.org/packages/3.0/bioc/html/customProDB.html>. Here, we provide the function `PrepareAnnotationGENCODE` to prepare the annotation from GENCODE. This function requires users to download GTF file, coding sequence and protein sequence FASTA files from GENCODE ftp <ftp://ftp.sanger.ac.uk/pub/>

`gencode/Gencode_human/`. Users should use the same version of annotations through the same project analysis. All the annotations are saved to a specified directory for latter use.

```
> library(proBAMr)

> gtfFile <- system.file("extdata", "test.gtf", package="proBAMr")
> CDSfasta <- system.file("extdata", "coding_seq.fasta", package="proBAMr")
> pepfasta <- system.file("extdata", "pro_seq.fasta", package="proBAMr")
> annotation_path <- tempdir()
> PrepareAnnotationGENCODE(gtfFile, CDSfasta, pepfasta,
+                           annotation_path, dbsnp=NULL,
+                           splice_matrix=FALSE, COSMIC=FALSE)
```

2.2 Preparing PSMs table

After preparing all the annotation files, the R package *pepXMLTab* is used to extract confident PSMs and related information from pepXML files. Other tools are also applicable at this step, as long as it generates similar tabular files, as shown below.

```
> passedPSM <- read.table(system.file("extdata", "passedPSM.tab",
+                            package="proBAMr"), sep='\t', header=TRUE)
> passedPSM[1:3, ]

          spectrum
1 00463_H12_P003361_B00L_A00_R1.9484.9484.2
2 00463_H12_P003361_B00L_A00_R1.9501.9501.2
3 00463_H12_P003361_B00L_A00_R1.9526.9526.2

          spectrumNativeID start_scan end_scan
1 controllerType=0 controllerNumber=1 scan=9484        9484      9484
2 controllerType=0 controllerNumber=1 scan=9501        9501      9501
3 controllerType=0 controllerNumber=1 scan=9526        9526      9526

  precursor_neutral_mass assumed_charge index retention_time_sec
1           1945.011                 2   1604       5941.112
2           1945.019                 2   1614       5951.951
3           1945.016                 2   1631       5963.760

  hit_rank peptide peptide_prev_aa peptide_next_aa
1         1 VNPTVFFDIAVDGEPLGR             M             V
2         1 VNPTVFFDIAVDGEPLGR             M             V
3         1 VNPTVFFDIAVDGEPLGR             M             V

          protein
1 ENST00000415933.1|ENSG00000196262.9|OTTHUMG00000023687.5|OTTHUMT00000341788.1|PPIA-007|PPIA|12
2 ENST00000415933.1|ENSG00000196262.9|OTTHUMG00000023687.5|OTTHUMT00000341788.1|PPIA-007|PPIA|12
3 ENST00000415933.1|ENSG00000196262.9|OTTHUMG00000023687.5|OTTHUMT00000341788.1|PPIA-007|PPIA|12

  num_tot_proteins calc_neutral_pep_mass massdiff num_tol_term
1                  4            1944.995 -0.01667663          2
2                  4            1944.995 -0.02461120          2
```

```

3           4           1944.995 -0.02192565           2
  num_missed_cleavages num_matched_ions tot_num_ions      mvh
1             0            21            31 46.20442
2             0            26            31 60.50605
3             0            22            31 52.51659
  mzFidelity    xcorr modification NTT
1 81.97849 4.276119      <NA> 1
2 104.25397 5.334539     <NA> 1
3 86.18693 5.057394     <NA> 1

```

2.3 Generate SAM file using PSMtab2SAM

The function PSMtab2SAM first finds the peptide location in protein sequences, then maps the coding sequence of the peptide back to the genome according to the annotation.

```

> load(system.file("extdata/Gencode", "exon_anno.RData", package="proBAMr"))
> load(system.file("extdata/Gencode", "proseq.RData", package="proBAMr"))
> load(system.file("extdata/Gencode", "procodingseq.RData", package="proBAMr"))
> options(stringsAsFactors=FALSE)
> passedPSM <- read.table(system.file("extdata", "passedPSM.tab",
+   package="proBAMr"), sep='\t', header=TRUE)
> outfile <- paste(tempdir(), '/test.sam', sep=' ')
> PSMtab2SAM(passedPSM, XScolumn='mvh', exon, proteinseq,
+   procodingseq, outfile)
> SAM <- read.table(file=outfile, sep='\t')
> dim(SAM)

```

```
[1] 40 21
```

```
> SAM[20:27, ]
```

	V1	V2	V3	V4	V5
20	00463_H12_P003361_BOOL_A00_R1.0.1.7307	16	chr11	65622810	255
21	00463_H12_P003361_BOOL_A00_R1.0.1.7350	0	chr7	44839340	255
22	00463_H12_P003361_BOOL_A00_R1.0.1.7441	0	chr7	44836381	255
23	00463_H12_P003361_BOOL_A00_R1.0.1.7457	0	chr7	44836381	255
24	00463_H12_P003361_BOOL_A00_R1.0.1.7898	16	chr5	133509648	255
25	00463_H12_P003361_BOOL_A00_R1.0.1.7915	16	chr5	133509648	255
26	00463_H12_P003361_BOOL_A00_R1.0.1.7933	16	chr5	133509648	255
27	00463_H12_P003361_BOOL_A00_R1.0.1.7952	16	chr1	26230237	255
	V6	V7	V8	V9	
20	42M	*	0	0	
21	45M	*	0	0	
22	12M2453N24M	*	0	0	
23	12M2453N24M	*	0	0	
24	51M	*	0	0	

```

25      51M * 0 0
26      51M * 0 0
27      39M * 0 0
28
29          V10  V11    V12
30      CAAAGGGCTTGCCTCCAGGGAGATGACGGCACTGCCCCCCAG * XA:Z:0
31      TCCATCTATGGGGAGAAATTGAAAGATGAGAACATTCTCATCCTAAAG * XA:Z:0
32          GTCTCCTTGAGCTGTTGCAGACAAGGTCCCAAAG * XA:Z:0
33          GTCTCCTTGAGCTGTTGCAGACAAGGTCCCAAAG * XA:Z:0
34  TTTGGCAATTCCACATCAACTCAAATATCTCTCCATCAGAACTCTGCAA * XA:Z:0
35  TTTGGCAATTCCACATCAACTCAAATATCTCTCCATCAGAACTCTGCAA * XA:Z:0
36  TTTGGCAATTCCACATCAACTCAAATATCTCTCCATCAGAACTCTGCAA * XA:Z:0
37          CCGAGGGCTGAGAATCAGCTCAAAAGCCTGGCCTGAGGC * XA:Z:0
38
39          V13  V14      V15     V16      V17    V18
40  NH:i:1 XL:i:1   XP:Z:LGGSAVISLEGKPL XC:i:2 XS:f:30.6722 XM:Z:-
41  NH:i:1 XL:i:1   XP:Z:SIYGEKFEDENFILK XC:i:2 XS:f:22.4909 XM:Z:-
42  NH:i:1 XL:i:1   XP:Z:VSFELFADKVPK XC:i:2 XS:f:21.321 XM:Z:-
43  NH:i:1 XL:i:1   XP:Z:VSFELFADKVPK XC:i:2 XS:f:25.2581 XM:Z:-
44  NH:i:1 XL:i:1   XP:Z:LQSSDGEIFEVDEIAK XC:i:2 XS:f:30.9829 XM:Z:-
45  NH:i:1 XL:i:1   XP:Z:LQSSDGEIFEVDEIAK XC:i:2 XS:f:42.8235 XM:Z:-
46  NH:i:1 XL:i:1   XP:Z:LQSSDGEIFEVDEIAK XC:i:2 XS:f:33.9951 XM:Z:-
47  NH:i:1 XL:i:1   XP:Z:ASGQAFELILSPR XC:i:2 XS:f:23.4655 XM:Z:-
48
49          V19  V20    V21
50  XN:i:0 XT:i:2 XG:Z:N
51  XN:i:1 XT:i:2 XG:Z:N
52  XN:i:1 XT:i:2 XG:Z:N
53  XN:i:1 XT:i:2 XG:Z:N
54  XN:i:0 XT:i:2 XG:Z:N
55  XN:i:0 XT:i:2 XG:Z:N
56  XN:i:0 XT:i:2 XG:Z:N
57  XN:i:0 XT:i:2 XG:Z:N

```

2.4 Convert SAM file to BAM and index

Add the header to the SAM file. Converted them to the binary BAM files using samtools <http://samtools.sourceforge.net/>. Sort and index them for fast access.

The bullet list below summarizes the steps after the SAM file been generated.

```

> paste('cat header test.sam > test_header.sam')
[1] "cat header test.sam > test_header.sam"

> paste('samtools view -S -b test_header.sam > test_header.bam')
[1] "samtools view -S -b test_header.sam > test_header.bam"

> paste('samtools sort test_header.bam > test_header_sort')

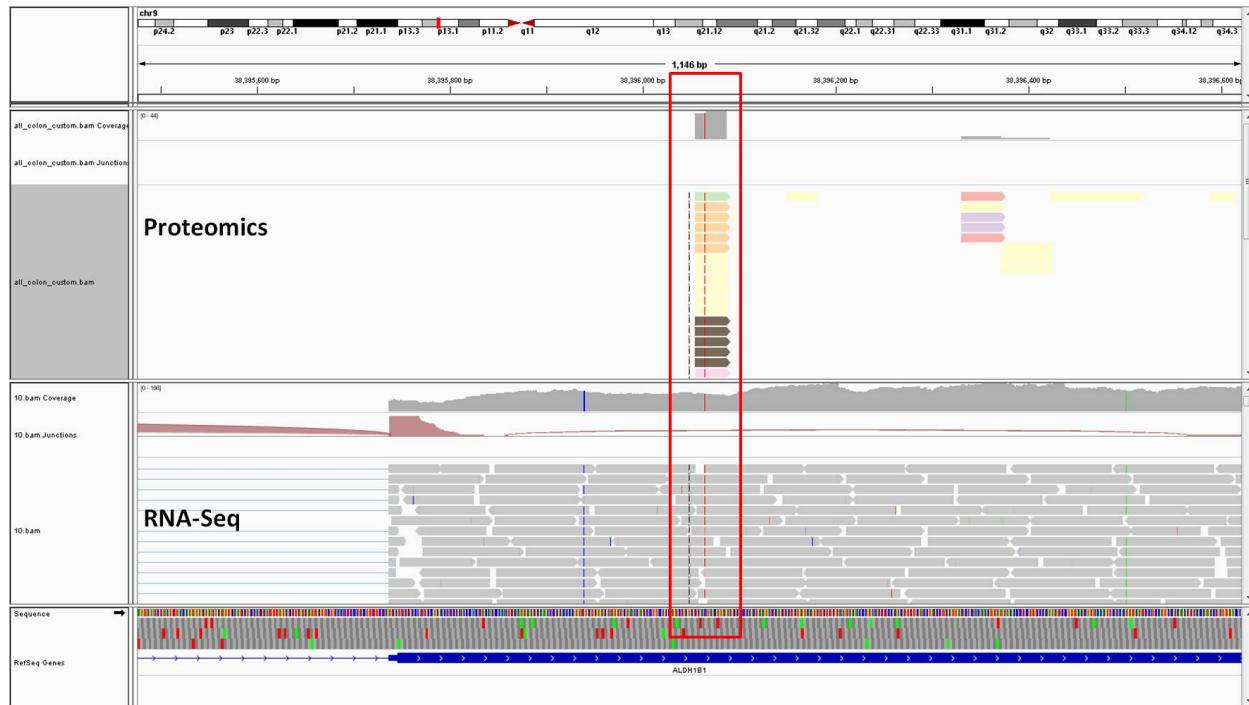
```

```
[1] "samtools sort test_header.bam > test_header_sort"
> paste('samtools index test_header_sort')
[1] "samtools index test_header_sort"
```

2.5 Visualize proteomics data in IGV

The proBAM files can be visualized in IGV directly. Furthermore, users can co-visualize their proteomics data with the paired genomics/transcriptomics data, as shown in Fig 2.

Figure 2: IGV snapshot of a homozygous mutation in gene ALDH1B1 in both proteomics and RNA-Seq data (inside read box)



3 Analyzing proBAM files using R package *proBAMtools*

This R package, *proBAMtools*, is designed to perform various analyses based on the proBAM files, including tools for proteomics data interpretation, assembly, and integration. This document provides a step by step tutorial using a toy genome, which only contains two genes.

3.1 Proteomics data interpretation

3.1.1 Summarize proteomics identifications from a proBAM file

The function `SummarizeproBAM` is used to summarize the proteomics identifications from a proBAM file. In addition to the number of identifiable spectra, PSMs and unique peptide, it is easy to identify genomic regions from which the peptides are derived with the genomic sequence mapping information available in the proBAM files. The function calculates the number of peptides mapped to unique or multiple genomic locations. And for peptides mapped to unique genomic locations, the function also provides the number of within exon peptides and exon-exon junction peptides.

```
> library(proBAMtools)

> proBAMFile <- system.file("extdata", "test.bam", package="proBAMtools")
> glan_proBAM <- readproBAM(proBAMFile)
> glan_proBAM

GAlignments object with 146 alignments and 15 metadata columns:
  seqnames strand      cigar      qwidth      start      end
  <Rle>   <Rle>   <character> <integer> <integer> <integer>
[1] chr12     +        27M       27  25357166 25357192
[2] chr12     +        27M       27  25357166 25357192
[3] chr12     +        27M       27  25357166 25357192
[4] chr12     +        27M       27  25357166 25357192
[5] chr12     +        27M       27  25357166 25357192
...
[142] chr12     - 12M17861N18M       30  25380335 25398225
[143] chr12     - 12M17861N18M       30  25380335 25398225
[144] chr12     -        54M       54  25398262 25398315
[145] chr12     -        33M       33  25398271 25398303
[146] chr12     -        33M       33  25398271 25398303
  width      njunc  |
  <integer> <integer> |
[1]      27        0  |
[2]      27        0  |
[3]      27        0  |
[4]      27        0  |
[5]      27        0  |
...
[142] 17891        1  |
[143] 17891        1  |
[144]    54        0  |
[145]    33        0  |
[146]    33        0  |
  qname      flag
  <character> <integer>
[1] klc_100908x_PH_P10_DLD_1_A12.0.1.8807        0
```

```

[2]          mam_072308x_PH_P8_HCT_116_B12.0.1.7396      0
[3] klc_070108x_PH_P7_HCT_15_C12_080712110959.0.1.7206      0
[4]          klc_080408x_PH_P8_HT_29_E12.0.1.7619      0
[5]          klc_080408x_PH_P8_LS174_T_F12.0.1.7883      0
...
[142]          ...          ...
[143]          klc_090308x_PH_P7a_SW480_J05.0.1.3992      16
[144]          mam_072308x_PH_P8_HCT_15_C11.0.1.6201      16
[145]          klc_082108x_PH_P7a_CaCo2_G15_1.0.1.4080      16
[146]          klc_090308x_PH_P7a_SW480_J15.0.1.4920      16
          seq
          <DNAStringSet>
[1]          GAGCTAGAACGCTTGTACTTCCTTAGG
[2]          GAGCTAGAACGCTTGTACTTCCTTAGG
[3]          GAGCTAGAACGCTTGTACTTCCTTAGG
[4]          GAGCTAGAACGCTTGTACTTCCTTAGG
[5]          GAGCTAGAACGCTTGTACTTCCTTAGG
...
[142]          ...          ...
[143]          CCTGTAGGAATCCTCTATTGTTGGATCATA
[144]          CAAGGCACCTTGCTACGTCACCAGCTCCAACCTACCACAAGTTATTCAGT
[145]          CTTGCCTACGCCACCAGCTCCAACCTACCACAAG
[146]          CTTGCCTACGCCAACAGCTCCAACCTACCACAAG
          NH          XL          XP          XR
          <integer> <integer> <character> <character>
[1]          1          1          ELEALYFLR          ELEALYFLR
[2]          1          1          ELEALYFLR          ELEALYFLR
[3]          1          1          ELEALYFLR          ELEALYFLR
[4]          1          1          ELEALYFLR          ELEALYFLR
[5]          1          1          ELEALYFLR          ELEALYFLR
...
[142]          ...          ...          ...
[143]          6          1          YDPTIEDSYR          YDPTIEDSYR
[144]          1          1          TEYKLVVVGAGDVGKSAL          TEYKLVVVGAGGVGKSAL
[145]          4          1          LVVVGAGGVGK          LVVVGAGGVGK
[146]          1          1          LVVVGAVGVGK          LVVVGAGGVGK
          XS          XQ          XC          XA          XM
          <numeric> <logical> <integer> <character> <character>
[1] 39.1477 <NA>          2          0          NA
[2] 39.2369 <NA>          2          0          NA
[3] 45.3205 <NA>          2          0          NA
[4] 50.0433 <NA>          2          0          NA
[5] 52.9704 <NA>          2          0          NA
...
[142] 49.2137 <NA>          2          0          NA

```

```

[143] 32.8392 <NA> 2 0 NA
[144] 41.6893 <NA> 4 0 NA
[145] 49.9949 <NA> 2 0 NA
[146] 45.6955 <NA> 2 0 NA
      XN XT XG
<integer> <integer> <character>
[1] 0 2 N
[2] 0 2 N
[3] 0 2 N
[4] 0 2 N
[5] 0 2 N
...
[142] 0 2 N
[143] 0 2 N
[144] 2 0 V
[145] 0 2 N
[146] 0 2 V
-----
seqinfo: 217 sequences from an unspecified genome

```

> *SummarizeproBAM(glan_proBAM)*

Spectra	
146	
PSM	
146	
Distinct_peptide	
14	
Distinct_peptide_seq	
12	
Peptide map to unique genomic location	
12	
Peptides within exon	
9	
Peptides spanning exons	
3	
Peptides spanning 2 exons	
3	
Peptides spanning 3 exons	
NA	
Peptides spanning > 3 exons	
0	
Peptide map to 2 genomic location	
NA	
Peptide map to 3 genomic location	
NA	

```

Peptide map to 4-10 genomic location
                                NA
Peptide map to >10 genomic location
                                NA

```

3.1.2 Sequence coverage analysis of proteomics data

While existing proteomics data formats only report sequence coverage for individual proteins, the proBAM format contains genome information, thereby allowing *proBAMtools* to provide sequence coverage reports at the genome, chromosome, and individual gene levels. The function *CDScov* is used to calculate sequence coverage at the whole genome or chromosome level.

```

> txdb <- loadDb(system.file("extdata/anno_ensembl", "txdb.sqlite",
+ package="proBAMtools"))
> CDScov(glan_proBAM, txdb, bychr=FALSE)

```

```
[1] 0.1579892
```

```
> CDScov(glan_proBAM, txdb, bychr=TRUE)
```

chr1	chr2	chr3	chr4	chr5	chr6	chr7
NaN	NaN	NaN	NaN	NaN	NaN	NaN
chr8	chr9	chr10	chr11	chr12	chr13	chr14
NaN	NaN	NaN	NaN	0.1579892	NaN	NaN
chr15	chr16	chr17	chr18	chr19	chr20	chr21
NaN	NaN	NaN	NaN	NaN	NaN	NaN
chr22	chrX	chrY				
NaN	NaN	NaN				

And function *GeneCDScov* is provided to calculate the sequence coverage for individual genes.

```
> GeneCDScov(glan_proBAM, txdb)
```

```
ENSG00000133703 ENSG00000205707
0.33474576      0.02288136
```

3.2 Proteomics data assembly

3.2.1 Parsimony

In a typical proteomics study, parsimony analysis is usually used to derive a minimal protein list that is sufficient to explain the observed peptide identification. We provide a function *proBAM2pepBAM* to convert a PSMs based proBAM file to the peptide based proBAM file.

```

> pepBAM <- proBAM2pepBAM(glan_proBAM)
> pepBAM

```

GAlignments object with 16 alignments and 13 metadata columns:

	seqnames	strand	cigar	qwidth	start	end
	<Rle>	<Rle>	<character>	<integer>	<integer>	<integer>
[1]	chr12	+	27M	27	25357166	25357192
[2]	chr12	-	33M15702N3M	36	25362813	25378550
[3]	chr12	-	33M15702N3M	36	25362813	25378550
[4]	chr12	-	30M	30	25368462	25368491
[5]	chr12	-	30M	30	25368462	25368491
...
[12]	chr12	-	12M17861N18M	30	25380335	25398225
[13]	chr12	-	12M17861N18M	30	25380335	25398225
[14]	chr12	-	54M	54	25398262	25398315
[15]	chr12	-	33M	33	25398271	25398303
[16]	chr12	-	33M	33	25398271	25398303
	width	njunc		qwidth	njunc	flag
	<integer>	<integer>		<integer>	<integer>	<integer>
[1]	27	0		27	0	0
[2]	15738	1		36	1	16
[3]	15738	1		36	1	16
[4]	30	0		30	0	16
[5]	30	0		30	0	16
...
[12]	17891	1		30	1	16
[13]	17891	1		30	1	16
[14]	54	0		54	0	16
[15]	33	0		33	0	16
[16]	33	0		33	0	16
	seq	<character>				
[1]		GAGCTAGAAGCTTGTACTTCCTTAGG				
[2]		TCGAACTAATGTATAGAAGGCATCATCAACACCCCTG				
[3]		TCGAACTAATGTATAGAAGGCATCATCAACACCCCTG				
[4]		CCTCACCAATGTATAAAAAGCATCCTCCAC				
[5]		CCTCACCAATGTATAAAAAGCATCCTCCAC				
...		...				
[12]		CCTGTAGGAATCCTCTATTGTTGGATCATA				
[13]		CCTGTAGGAATCCTCTATTGTTGGATCATA				
[14]	CAAGGCACTCTGCCTACGTCACCAGCTCCAACCTACCAAGTTATATTCAAGT					
[15]		CTTGCCCTACGCCACCAGCTCCAACCTACCAAG				
[16]		CTTGCCCTACGCCAACAGCTCCAACCTACCAAG				
	XA	NH	XP		XR	
	<character>	<integer>	<character>	<character>	<character>	
[1]	0	1	ELEALYFLR		ELEALYFLR	
[2]	0	1	QGVDDAFYTLVR		QGVDDAFYTLVR	
[3]	0	1	QGVDDAFYTLVR		QGVDDAFYTLVR	

```

[4]          0      4      VEDAFYTLVR      VEDAFYTLVR
[5]          0      4      VEDAFYTLVR      VEDAFYTLVR
...
[12]         ...    ...
[12]          0      6      YDPTIEDSYR     YDPTIEDSYR
[13]          0      6      YDPTIEDSYR     YDPTIEDSYR
[14]          0      1  TEYKLVVVGAGDVGKSAL  TEYKLVVVGAGGVGKSAL
[15]          0      4      LVVVGAGGVGK     LVVVGAGGVGK
[16]          0      1      LVVVGAVGVGK     LVVVGAGGVGK

          XC           XM           XN           XT           XG
<integer> <character> <integer> <integer> <character>
[1]          2          -          0          2          N
[2]          2          -          0          2          N
[3]          2 1;111.0320285114 0          2          N
[4]          2          -          0          1          N
[5]          2          -          0          2          N
...
[12]         ...    ...
[12]          2          -          0          2          N
[13]          2          -          0          1          N
[14]          4          -          2          0          V
[15]          2          -          0          2          N
[16]          2          -          0          2          V
-----
seqinfo: 1 sequence from an unspecified genome; no seqlengths

```

In function `pepBAM_parsimony`, a previously published parsimony procedure is used to generate a minimal list of identified proteins or genes. The procedure relies on bipartite graph modeling. What distinguishes the proBAM approach from the previous approach is the utilization of genomic location information in the construction of the bipartite graph. Specifically, the proBAM-based approach can relate peptides to different types of genomic elements such as exons, transcript isoforms, or genes based on their overlapping relationship on the genome. Consequently, inference can be made at both protein level and gene level. Moreover, inference can be made based on different versions of genome annotations.

```
> #gene level parsimony  
> cdsByGe <- cdsBy(txdb, "gene", use.names=FALSE)  
> gegp <- pepBAM_parsimony(pepBAM, cdsByGe)  
> gegp
```

```
Group
[1,] "ENSG00000205707"
[2,] "ENSG00000133703"
Peptides
[1,] "ELEALYFLR 2 -_pep"
[2,] "QGVDDAFYTLVR 2 -_pep:QGVDDAFYTLVR 2 1;111.0320285114_pep:VEDAFYTLVR 2 -_pep:QAQDLAR 1 1;11
#inGroup #ofPeptide
```

```

[1,] "1"      "1"
[2,] "1"      "13"

> #protein level parsimony
> cdsByTx <- cdsBy(txdb, "tx", use.names=TRUE)
> progp <- pepBAM_parsimony(pepBAM, cdsByTx)
> progp

  Group
[1,] "ENST00000381356:ENST00000556351:ENST00000556885:ENST00000556927:ENST00000557540"
[2,] "ENST00000311936"
[3,] "ENST00000256078"

  Peptides
[1,] "ELEALYFLR 2 -_pep"
[2,] "QGVDDAFYTLVR 2 -_pep:QGVDDAFYTLVR 2 1;111.0320285114_pep:QAQDLAR 1 1;111.0320285114_pep:DS
[3,] "VEDAFYTLVR 2 -_pep:QAQDLAR 1 1;111.0320285114_pep:DSEDVPMVLVGNKCDLPSR 3 7;147.0353996062;1
  #inGroup #ofPeptide
[1,] "5"      "1"
[2,] "1"      "12"
[3,] "1"      "11"

```

3.2.2 Generate spectra count table for proteomics data

proBAMtools generates count tables on the basis of genomic structure of the genes. Both spectral count and peptide count tables are provided at protein isoform- and gene-levels, respectively, and both overall counts and gene-specific or isoform-specific counts are reported. The overall spectral count for a gene or an isoform respectively sums up all PSMs associated with the gene or the isoform, whereas the gene-specific or isoform-specific spectral count, respectively, sums up only PSMs associated specifically with the gene or the isoform. The same conditions also apply to peptide counting. Count data are provided for individual genes and proteins. By integrating these data with the minimal protein or gene group lists resulting from the parsimony analysis, count tables can be generated for the confidently identified genes or proteins.

```

> ##gene level peptide count
> cdsByGe <- cdsBy(txdb, "gene", use.names=FALSE)
> probAMCount(pepBAM, cdsByGe)

```

	specific	overall
ENSG00000133703	8	15
ENSG00000205707	1	1

```

> ##gene level spectra count
> probAMCount(glan_probAM, cdsByGe)

```

	specific	overall
ENSG00000133703	78	139
ENSG00000205707	7	7

```

> ##protein level peptide count
> cdsByTx <- cdsBy(txdb, "tx", use.names=TRUE)
> proBAMCount(pepBAM, cdsByTx)

      specific overall
ENST00000256078      0     13
ENST00000311936      2     13
ENST00000381356      0      1
ENST00000553788      0      0
ENST00000554266      0      0
ENST00000555151      0      0
ENST00000555711      0      0
ENST00000556131      0      3
ENST00000556198      0      0
ENST00000556351      0      1
ENST00000556402      0      0
ENST00000556885      0      1
ENST00000556927      0      1
ENST00000557334      0      3
ENST00000557540      0      1

> ##protein level spectra count
> proBAMCount(glan_proBAM, cdsByTx)

      specific overall
ENST00000256078      0     85
ENST00000311936     54    124
ENST00000381356      0      7
ENST00000553788      0      0
ENST00000554266      0      0
ENST00000555151      0      0
ENST00000555711      0      0
ENST00000556131      0      3
ENST00000556198      0      0
ENST00000556351      0      7
ENST00000556402      0      0
ENST00000556885      0      7
ENST00000556927      0      7
ENST00000557334      0      3
ENST00000557540      0      7

```

3.3 Proteomics data integration

3.3.1 Switch annotation

Using genomic sequence as a common reference, proBAM provides a natural solution for proteomics data integration. PSMs align with the genome regardless of the protein database used in the pro-

teomics study, thereby allowing flexible switching between different gene annotation schemes. We developed a function `Switchanno`, that use a three-step procedure to switch gene annotation scheme for proBAM files, which includes: 1) Removing peptides with any region out of the CDSs of the new annotation scheme; 2) Removing peptides with any region inconsistent with the gene structure of the new scheme; 3) Removing peptides that are out of frame in the new scheme. Accordingly, proteomics datasets generated by searching against different databases can be converted to the same gene annotation scheme for integration.

```
> new_annotation_path <- system.file("extdata/anno_refseq/",
+ package="proBAMtools")
> Switchanno(glan_proBAM, new_annotation_path)
```

GAlignments object with 139 alignments and 15 metadata columns:

	seqnames	strand	cigar	qwidth	start	end
[1]	chr12	-	33M15702N3M	36	25362813	25378550
[2]	chr12	-	33M15702N3M	36	25362813	25378550
[3]	chr12	-	33M15702N3M	36	25362813	25378550
[4]	chr12	-	33M15702N3M	36	25362813	25378550
[5]	chr12	-	33M15702N3M	36	25362813	25378550
...
[135]	chr12	-	45M	45	25378647	25378691
[136]	chr12	-	39M	39	25378647	25378685
[137]	chr12	-	54M	54	25398262	25398315
[138]	chr12	-	33M	33	25398271	25398303
[139]	chr12	-	33M	33	25398271	25398303
	width	njunc				
	<integer>	<integer>				
[1]	15738	1				
[2]	15738	1				
[3]	15738	1				
[4]	15738	1				
[5]	15738	1				
...			
[135]	45	0				
[136]	39	0				
[137]	54	0				
[138]	33	0				
[139]	33	0				
	qname	flag				
	<character>	<integer>				
[1]	klc_080408x_PH_P8_CaCo2_G08.0.1.6578	16				
[2]	klc_080408x_PH_P8_CaCo2_G08.0.1.6645	16				
[3]	klc_080408x_PH_P8_CaCo2_G08.0.1.7102	16				
[4]	klc_080408x_PH_P8_CaCo2_G08.0.1.7110	16				
[5]	klc_082108x_PH_P7a_CaCo2_G08_1.0.1.6905	16				

```

...
[135]   klc_090308x_PH_P7a_SW480_J15.0.1.5465      16
[136]   mam_20081229x_P10_SW480_J03.0.1.5910      16
[137]   mam_072308x_PH_P8_HCT_15_C11.0.1.6201      16
[138] klc_082108x_PH_P7a_CaCo2_G15_1.0.1.4080      16
[139]   klc_090308x_PH_P7a_SW480_J15.0.1.4920      16
          seq
<DNAStringSet>
[1] TCGAACTAATGTATAGAACGGCATCATCAACACCCCTG
[2] TCGAACTAATGTATAGAACGGCATCATCAACACCCCTG
[3] TCGAACTAATGTATAGAACGGCATCATCAACACCCCTG
[4] TCGAACTAATGTATAGAACGGCATCATCAACACCCCTG
[5] TCGAACTAATGTATAGAACGGCATCATCAACACCCCTG
...
[135]   TTTATTCCTACTAGGACCATAGGTACATCTCAGAGTCCTAAC
[136]   TTTATTCCTACTAGGACCATAGGTACATCTCAGAGTC
[137] CAAGGCACTCTGCCTACGTACCAGCTCCAACCTACCACAAGTTTATTCAGT
[138]   CTTGCCTACGCCACCAGCTCCAACCTACCACAAG
[139]   CTTGCCTACGCCAACAGCTCCAACCTACCACAAG
      NH       XL           XP           XR
<integer> <integer> <character> <character>
[1]     1       1    QGVDDAFYTLVR  QGVDDAFYTLVR
[2]     1       1    QGVDDAFYTLVR  QGVDDAFYTLVR
[3]     1       1    QGVDDAFYTLVR  QGVDDAFYTLVR
[4]     1       1    QGVDDAFYTLVR  QGVDDAFYTLVR
[5]     1       1    QGVDDAFYTLVR  QGVDDAFYTLVR
...
[135]     ...     ...
[136]     ...     ...
[137]     ...     ...
[138]     ...     ...
[139]     ...     ...
      XS       XQ       XC       XA           XM
<numeric> <logical> <integer> <character> <character>
[1] 39.3926 <NA>     2       0       NA
[2] 41.2553 <NA>     2       0       NA
[3] 52.9468 <NA>     2       0 1;111.0320285114
[4] 45.3983 <NA>     2       0 1;111.0320285114
[5] 59.9704 <NA>     2       0       NA
...
[135]     ...     ...
[136]     ...     ...
[137]     ...     ...
[138]     ...     ...
[139]     ...     ...

```

```

      XN      XT      XG
<integer> <integer> <character>
[1]      0      2      N
[2]      0      2      N
[3]      0      2      N
[4]      0      2      N
[5]      0      2      N
...
[135]     1      2      N
[136]     0      2      N
[137]     2      0      V
[138]     0      2      N
[139]     0      2      V
-----
seqinfo: 24 sequences from an unspecified genome

```

```

> length(glan_proBAM)
[1] 146

> length(Switchanno(glan_proBAM, new_annotation_path))
[1] 139

```

3.3.2 Multiple data sets integration

proBAM files are generated from individual studies using *proBAMr* based on user-specified PSM FDRs. Because PSMs in all resulting proBAM files are mapped to the genome and thus aligned in the same coordinate system, they can be easily combined into one proBAM file. All PSMs in this combined proBAM file can be re-annotated according to a user-specified gene annotation scheme as described above.

4 Session Information

```
R version 3.1.1 (2014-07-10)
Platform: x86_64-unknown-linux-gnu (64-bit)
```

```
locale:
[1] LC_CTYPE=en_US.UTF-8      LC_NUMERIC=C
[3] LC_TIME=en_US.UTF-8       LC_COLLATE=en_US.UTF-8
[5] LC_MONETARY=en_US.UTF-8   LC_MESSAGES=en_US.UTF-8
[7] LC_PAPER=en_US.UTF-8      LC_NAME=C
[9] LC_ADDRESS=C              LC_TELEPHONE=C
[11] LC_MEASUREMENT=en_US.UTF-8 LC_IDENTIFICATION=C
```

```
attached base packages:
```

```
[1] stats4    parallel   stats      graphics  grDevices utils
[7] datasets  methods    base

other attached packages:
[1] proBAMtools_0.99.0      Matrix_1.2-2
[3] igraph_1.0.1            GenomicFeatures_1.18.7
[5] GenomicAlignments_1.2.2 Rsamtools_1.18.3
[7] Biostrings_2.34.1       XVector_0.6.0
[9] GenomicRanges_1.18.4    proBAMr_1.3.2
[11] AnnotationDbi_1.28.2   GenomeInfoDb_1.2.5
[13] Biobase_2.26.0          IRanges_2.0.1
[15] S4Vectors_0.4.0         BiocGenerics_0.12.1

loaded via a namespace (and not attached):
[1] base64enc_0.1-3     BatchJobs_1.6      BBmisc_1.9
[4] BiocParallel_1.0.3   biomaRt_2.22.0    bitops_1.0-6
[7] brew_1.0-6           checkmate_1.6.2  codetools_0.2-14
[10] DBI_0.3.1           digest_0.6.8     fail_1.2
[13] foreach_1.4.2       grid_3.1.1      iterators_1.0.7
[16] lattice_0.20-33    magrittr_1.5    RCurl_1.95-4.7
[19] RSQLite_1.0.0        rtracklayer_1.26.3 sendmailR_1.2-1
[22] stringi_0.5-5      stringr_1.0.0    tools_3.1.1
[25] XML_3.98-1.3        zlibbioc_1.12.0
```