

Introduction to *VariantAnnotation*

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1 Introduction

In this vignette we introduce methods in the *VariantAnnotation* package for the annotation and filtering of genetic variants. Variant location with respect to gene function can be determined as well as the amino acid coding changes for nonsynonymous variants. *VariantAnnotation* contains functions to access the SIFT and PolyPhen databases packages. The SIFT and PolyPhen methods predict the possible impact of amino acid coding changes on protein function (SIFT) or function and structure (PolyPhen).

To assist with data access and management the `readVcf` function parses data from Variant Call Format (VCF) files into a *summarizedExperiment* object. Filtering functions are available for subsetting of variants on physical location in the gene or membership in dbSNP.

2 Variant Location

As sample data we use the `GRanges` object, `variants`, that contains both real and faux variants. The ranges in `variants` represent the positions in the reference sequence that are altered in the variation. A single basepair replacement (SNP) has equal start and end positions and a width of length 1. A deletion or substitution will have a width of 1 or greater depending on the number of basepairs being deleted or substituted. Insertions are represented with start = end + 1 which results in a width of length 0.

Reference and variant alleles are represented as `DNAStringSets` in `refAllele` and `varAllele` columns. An insertion will have a missing value in the `refAllele` column and a deletion will have a missing value in the `varAllele` column.

```
> library(VariantAnnotation)
> data(variants)
> head(variants)

GRanges with 6 ranges and 3 elementMetadata values:
  seqnames      ranges strand |      refAllele
    <Rle>      <IRanges>  <Rle> |      <DNAStringSet>
 [1] chr1 [161003087, 161003087] * |      C
 [2] chr1 [ 84647761, 84647761] * |      C
 [3] chr1 [ 13133880, 13133881] * |      TC
 [4] chr1 [ 11326183, 11326182] * |
 [5] chr1 [105293754, 105293754] * |      A
 [6] chr1 [ 67478546, 67478546] * |      G
  varAllele      comments
  <DNAStringSet>      <character>
 [1]      T      rs1000050, SNP
 [2]      T      rs6576700 SNP
 [3]          rs59770105, deletion
```

```

[4]      AT    rs35561142, insertion
[5]  ATAAA rs10552169, substitution
[6]      A    rs11209026, SNP
---
seqlengths:
chr1 chr13 chr16 chr2
NA     NA     NA     NA

```

`locateVariants` returns a `DataFrame` with one row for each variant-transcript match. The `queryHits` column maps back to the variant in the original query. Location categories of ‘coding’, ‘UTR5’, ‘UTR3’, ‘intron’ and ‘intergenic’ are assigned based on the criteria in Table 1. Intergenic variants have two gene ID’s, representing the genes preceding and following the variant.

```

> library(TxDb.Hsapiens.UCSC.hg18.knownGene)
> txdb <- TxDb.Hsapiens.UCSC.hg18.knownGene
> loc <- locateVariants(variants, txdb)
> head(loc)

```

```

DataFrame with 6 rows and 4 columns
  queryHits      txID          geneID  Location
  <integer> <character> <CompressedCharacterList> <factor>
1         1      2966           4921   intron
2         1      2967           4921   intron
3         2      1660          58511   intron
4         2      1661          58511   intron
5         2      1662          58511   intron
6         3        NA 343068,401940 intergenic

```

Region	Details
coding	variant falls <i>within</i> a coding region
UTR5	variant falls <i>within</i> a 5’ untranslated region
UTR3	variant falls <i>within</i> a 3’ untranslated region
intron	variant falls <i>within</i> a transcript but not <i>within</i> a coding region
intergenic	variant does not fall <i>within</i> a transcript

Table 1: Variant Location

Multiple alignments for a single variant should be separate rows in the `query` object. Below is an example `snp` from dbSNP Build 130 which is aligned to different locations for different assemblies. Each alignment is entered as a separate row in the `singleSNP` object.

```

> singleSNP <- GRanges(seqnames="chr16",
+   ranges=IRanges(start=c(23004226, 22316999, 24133766),
+   width=c(1,1,1)),
+   varAllele=DNAStringSet("T"),
+   comments=c("Celera assembly", "HuRef assembly", "reference assembly"))
> singleSNP

GRanges with 3 ranges and 2 elementMetadata values:
  seqnames      ranges strand |      varAllele
  <Rle>      <IRanges> <Rle> | <DNAStringSet>

```

```

[1] chr16 [23004226, 23004226]      * |      T
[2] chr16 [22316999, 22316999]      * |      T
[3] chr16 [24133766, 24133766]      * |      T
      comments
      <character>
[1] Celera assembly
[2] HuRef assembly
[3] reference assembly
---
seqlengths:
chr16
NA

```

The result from `locateVariants` shows the Celera assembly location matching two transcripts in the same gene, both intron regions. The HuRef assembly falls in an intergenic region. The reference assembly hits two different transcripts in the same gene, both in the 3' UTR region.

```

> locateVariants(singleSNP, txdb)
DataFrame with 5 rows and 4 columns
  queryHits      txID          geneID    Location
  <integer> <character> <CompressedCharacterList> <character>
1        1      52091           57478     intron
2        1      52092           57478     intron
3        2       NA          1039,57478   intergenic
4        3      50256            5579     3'UTR
5        3      50259            5579     3'UTR

```

3 Amino Acid Coding

Amino acid coding changes can be determined for non-synonymous variants with `predictCoding`. This function identifies overlaps between the variants in `query` and the coding regions of a `TranscriptDb` object. Sequences for the reference codons are retrieved from either a `BSgenome` or fasta file. Codon sequences for the variants are constructed by substituting, inserting or deleting the `varAllele` values into the reference sequence. Amino acid codes are computed for the resulting variant codon sequence when the length is a multiple of 3. Examples of various coding situations are shown in Table 2. Positions of substitution, insertion or deletion in Table 2 are purely random and were chosen for example purposes only.

Type	refAllele	varAllele	refSeq	varSeq	AA coding of varSeq
substitution	G	T	aag	aaT	yes
substitution	G	TG	tga	tTGa	no
substitution	G	TGCG	gtc	TGCGtc	yes
insertion	"	G	cgg	Gggg	no
insertion	"	TTG	gaa	gaTTGa	yes
deletion	A	"	atc	tc	no
deletion	GGCCTA	"	acggcctaa	aca	yes

Table 2: Amino Acid Coding

Variants that fall within coding regions are in chromosomes 1, 2 and 16. We subset the `GRanges` of coding regions by transcripts to retain only the chromosomes of interest. Alternatively one could use the full `TranscriptDb`.

```

> library(BSgenome.Hsapiens.UCSC.hg18)
> cdsByTx <- cdsBy(txdb)
> grl <- keepSeqlevels(cdsByTx, c("chr1", "chr2", "chr16"))
> aaCoding <- predictCoding(variants, grl, seqSource=Hsapiens,
+   varAllele="varAllele")
> head(aaCoding)

DataFrame with 6 rows and 10 columns
  queryHits      txID      refSeq      varSeq      refAA
  <integer> <character> <DNAStringSet> <DNAStringSet> <AAStringSet>
1       6        2520        CGA        CAA        R
2       6        2521        CGA        CAA        R
3       6        2522        CGA        CAA        R
4       6        2524        CGA        CAA        R
5       7       10566        ACT        GCT        T
6       7       10567        ACT        GCT        T
  varAA  Consequence      cds_id  cds_name exon_rank
  <AAStringSet>    <factor> <integer> <character> <integer>
1       Q nonsynonymous     8820      NA        5
2       Q nonsynonymous     8821      NA        6
3       Q nonsynonymous     8822      NA        7
4       Q nonsynonymous     8823      NA        8
5       A nonsynonymous     8824      NA        9
6       A nonsynonymous     8825      NA       10

```

The result is a DataFrame containing only the variants in coding regions. Columns include queryHits, txID, refSeq, varSeq, refAA, varAA, Consequence and any metadata columns that were present in the subject.

The result has one row for each transcript matched by a variant. The queryHits column is the map back to the variants in the original query. Multiple rows are returned for variants that match multiple transcripts. For example, the variant in the sixth row of `variants` fell within a coding region and matched four transcripts. The first four rows of the aaCoding result are the data for this variant.

The Consequence column indicates ‘synonymous’ or ‘nonsynonymous’ for variants for which the codons could be translated. If translation was not possible the Consequence was considered a ‘frameshift’. Variants for which no amino acid codes were computed have a missing in value in the varAA column.

```

> aaCoding[width(aaCoding$varAA) == 0,]

DataFrame with 8 rows and 10 columns
  queryHits      txID      refSeq      varSeq      refAA
  <integer> <character> <DNAStringSet> <DNAStringSet> <AAStringSet>
1       10        51251        CAG        CACG        Q
2       11        51251        CAG      CACTGGG        Q
3       13        51251        GCC          C        A
4       15        51249        CGG        TTGG        R
5       15        51250        CGG        TTGG        R
6       15        51251        CGG        TTGG        R
7       15        51252        CGG        TTGG        R
8       15        51253        CGG        TTGG        R
  varAA  Consequence      cds_id  cds_name exon_rank
  <AAStringSet>    <factor> <integer> <character> <integer>
1 frameshift      8822      NA        7

```

2	frameshift	8822	NA	7
3	frameshift	8822	NA	7
4	frameshift	8830	NA	5
5	frameshift	8821	NA	6
6	frameshift	8822	NA	7
7	frameshift	8823	NA	8
8	frameshift	8824	NA	9

4 Filtering

Identifying subsets of variants that meet a specific criteria can be accomplished using the `dbSNPfilter` or `regionFilter` functions. Filters are created with the appropriate reference object such as a `SNPlocs` or `TranscriptDb` object.

```
> library("SNPlocs.Hsapiens.dbSNP.20090506")
> ## create filter to identify variants present in dbSNP
>.snpFilt <- dbSNPFilter("SNPlocs.Hsapiens.dbSNP.20090506")
> ## create filter to identify variants present in introns
> regionFilt <- regionFilter(txdb, region="intron")
> regionFilt

function (x, subset = TRUE)
{
  res <- function (x)
  {
    loc <- locateVariants(x, txdb)
    res <- logical(length(x))
    res[unique(loc$queryHits[loc$Location %in% region])] <- TRUE
    res
  }(x)
  VAFFilterResult(x, res, "regionFilter", subset)
}
<environment: 0x104d13b8>
attr(,"name")
"regionFilter"
attr(,"class")
[1] "VAFfilter"
attr(,"class")attr(,"package")
[1] "VariantAnnotation"
```

When filtering data with `dbSNPFilter` or `regionsFilt` two options are available for the return object. If the argument `subset=TRUE` (default) a `GRanges` is returned that contains only the records that passed the filter. If `subset=FALSE`, the result is a `VAFfilterResult` object. This object contains a logical vector indicating which records passed the filter and some stats on the number of records removed and what filter was applied.

```
> ## return the subset of records that passed the filter
>.snpResult <-.snpFilt(variants)
> .snpResult

GRanges with 6 ranges and 3 elementMetadata values:
  seqnames      ranges strand |      refAllele
          1       8822     + |
```

```

<Rle>           <IRanges> <Rle> | <DNAStringSet>
[1] chr1 [161003087, 161003087] * | C
[2] chr1 [ 84647761, 84647761] * | C
[3] chr1 [ 67478546, 67478546] * | G
[4] chr2 [233848107, 233848107] * | A
[5] chr16 [ 49303427, 49303427] * | C
[6] chr16 [ 49314041, 49314041] * | G
      varAllele   comments
<DNAStringSet>   <character>
[1] T rs1000050, SNP
[2] T rs6576700 SNP
[3] A rs11209026, SNP
[4] G rs2241880, SNP
[5] T rs2066844, SNP
[6] C rs2066845, SNP
---
seqlengths:
chr1 chr13 chr16 chr2
NA     NA     NA     NA

> metadata(snpResult)

$name
"dbSNPFilter"

$stats
  Name Input Passing  Op
1 dbSNPFilter    17      6 <NA>

> ## return a VAFilterResult object
> snpFilt(variants, subset=FALSE)

class: VAFilterResult
name: dbSNPFilter
output: TRUE TRUE ... FALSE FALSE
stats:
  Name Input Passing  Op
1 dbSNPFilter    17      6 <NA>

Filters can be combined to act in unison,

# return a subset of data passing both filters
> combofilt <- compose(snpFilt, regionFilt)
> combofilt(variants)
region= intron
GRanges with 2 ranges and 3 elementMetadata values
  seqnames      ranges strand |      refAllele      varAllele
  <Rle>           <IRanges> <Rle> | <DNAStringSet> <DNAStringSet>
[1] chr1 [161003087, 161003087] * | C          T
[2] chr1 [ 84647761, 84647761] * | C          T
      comments
      <character>
[1] rs1000050, SNP

```

```
[2] rs6576700 SNP

seqlengths
chr1 chr13 chr16 chr2
NA     NA     NA     NA
```

5 Variant Call Format (VCF) files

The `readVcf` function reads VCF files and parses them into a `SummarizedExperiment` object. When the compressed VCF file has a tabix file index a `GRanges` can be used to specify a subset of ranges. The sample VCF in `VariantAnnotation` is not compressed so in this example we compress on the fly, create a tabix index and retrieve a subset of ranges.

```
> ## compress vcf file , create index
> vcfFile <- system.file("extdata", "ex1.vcf", package="VariantAnnotation")
> from <- vcfFile
> to <- tempfile()
> compressVcf <- bgzip(from, to)
> idx <- indexTabix(compressVcf, "vcf")
> tab <- TabixFile(compressVcf, idx)
> head <- headerTabix(tab)
> ## select ranges with param
> gr <- GRanges(seqnames="1",
+                 ranges=IRanges(start=0, end=100000))
> vcf <- readVcf(tab, param=gr)
```

All data are parsed according to the header information in the VCF file. The header is stored as a `SimpleList` in the `exptData` slot and can be viewed by querying the elements of the list.

```
> vcf

class: SummarizedExperiment
dim: 7 2
exptData(1): HEADER
assays(3): GT GQ DP
rownames(7): . . . . .
rowData values names(5): REF ALT QUAL FILTER DP
colnames(2): A B
colData names(1): Samples

> exptData(vcf)[["HEADER"]]

SimpleDataFrameList of length 4
names(4): META FILTER FORMAT INFO

> exptData(vcf)[["HEADER"]][["META"]]

DataFrame with 1 row and 1 column
      Value
<character>
fileformat    VCFv4.0
```

The eight required fields of the VCF file are parsed into a GRanges object in the rowData slot. The ALT alleles are provided as a DataFrameList to accomodate for multiple values per variant. All fields in the INFO column are parsed into individual elementMetadata columns.

```
> exptData(vcf)[["HEADER"]][["FILTER"]]
DataFrame with 1 row and 1 column
  Description
  <character>
q10 Quality below 10

> exptData(vcf)[["HEADER"]][["INFO"]]
DataFrame with 1 row and 3 columns
  Number      Type Description
  <character> <character> <character>
DP          1     Integer Total Depth

> rowData(vcf)

GRanges with 7 ranges and 5 elementMetadata values:
  seqnames      ranges strand |           REF
  <Rle>    <IRanges> <Rle> | <DNAStringSet>
.       1 [100, 100]   * |           G
.       1 [110, 110]   * |           C
.       1 [120, 120]   * |           T
.       1 [130, 132]   * |         GAA
.       1 [140, 141]   * |         GT
.       1 [150, 154]   * |        TAAAA
.       1 [160, 161]   * |         TA
  ALT      QUAL     FILTER      DP
  <CompressedSplitDataFrameList> <numeric> <character> <integer>
.       ##### 1806     q10      35
.       ##### 1792     PASS      32
.       ##### 628      q10      21
.       ##### 1016     PASS      22
.       ##### 727      PASS      30
.       ##### 246      PASS      10
.       ##### 246      PASS      10
---
seqlengths:
  1
NA

> ## ALT as a DNAStringSetList
> head(as.list(values(rowData(vcf)))[["ALT"]]), 4)

$`1`
DataFrame with 1 row and 1 column
  ALT
  <DNAStringSet>
1      C
```

```

$`2`
DataFrame with 0 rows and 1 column

$`3`
DataFrame with 1 row and 1 column
      ALT
<DNAStringSet>
1       G

$`4`
DataFrame with 2 rows and 1 column
      ALT
<DNAStringSet>
1       G
2     GTA

The genotype data described in the FORMAT fields are parsed into matrices in the assays slot.

> exptData(vcf)[["HEADER"]][["FORMAT"]]

DataFrame with 3 rows and 3 columns
  Number      Type Description
<character> <character> <character>
GT          1    String   Genotype
GQ          1    Integer  Genotype Quality
DP          1    Integer   Read Depth

> assays(vcf)

SimpleList of length 3
names(3): GT GQ DP

> head(assays(vcf)$GT)

  A      B
. "1/1"  "1/1"
. "0/0"  "0/0"
. "0|1|1" "0|1|1"
. "1/2"  "1/2"
. "0/1"  "0/1"
. "1/2"  "1/2"

```

6 SIFT and PolyPhen Databases

SIFT (Sorting Intolerant From Tolerant) and PolyPhen (Polymorphism Phenotyping) are methods for predicting the impact of amino acid substitution on a human protein. The SIFT method uses sequence homology and the physical properties of amino acids to make predictions about protein function. PolyPhen uses sequence-based features and structural information characterizing the substitution to make predictions about the structure and function of the protein. Both methods offer interactive tools and documentation on their web sites,

<http://sift.bii.a-star.edu.sg/> <http://genetics.bwh.harvard.edu/pph2/>

Collated predictions for specific dbSNP builds are available as downloads from the SIFT and PolyPhen web sites. These results have been stored as sqlite databases and made available through the Bioconductor

packages *SIFT.Hsapiens.dbSNP132.db* and *PolyPhen.Hapiens.dbSNP131.db*. The packages are designed to be searched by rsid.

```
> library(SIFT.Hsapiens.dbSNP132)
> library(PolyPhen.Hsapiens.dbSNP131)
> ## rsids in the package
> head(keys(SIFT.Hsapiens.dbSNP132))

[1] "rs10000692" "rs10001580" "rs10002700" "rs10003238" "rs10003369"
[6] "rs10004"

> ## list available columns
> cols(SIFT.Hsapiens.dbSNP132)

[1] "RSID"          "PROTEINID"      "AACHANGE"      "METHOD"
[5] "AA"            "PREDICTION"     "SCORE"        "MEDIAN"
[9] "POSITIONSEQS"  "TOTALSEQS"    NA

> rsids <- c("rs2142947", "rs8692231", "rs3026284")
> subst <- c("RSID", "PREDICTION", "SCORE", "AACHANGE", "PROTEINID")
> select(SIFT.Hsapiens.dbSNP132, keys=rsids, cols=subst)

  RSID PROTEINID AACHANGE PREDICTION SCORE
1 rs2142947 NP_001019832 F430L TOLERATED 1.00
2 rs2142947 NP_001019832 F430L TOLERATED 0.74
3 rs2142947 NP_001019832 F430L TOLERATED 0.72
4 rs2142947 NP_001019832 F430L TOLERATED 1
5 rs8692231  XP_537288  P37R NOT SCORED <NA>
6 rs8692231  XP_537288  P37R NOT SCORED <NA>
7 rs8692231  XP_537288  P37R NOT SCORED <NA>
8 rs8692231  XP_537288  P37R NOT SCORED <NA>
9 rs3026284   NP_004920  G202D DELETERIOUS 0.03
10 rs3026284  NP_004920  G202D TOLERATED 1.00
11 rs3026284  NP_004920  G202D DELETERIOUS 0.00
12 rs3026284  NP_004920  G202D TOLERATED 1

> select(PolyPhen.Hsapiens.dbSNP131, keys="rs3026284",
+       cols=c("TRAININGSET", "PREDICTION", "PPH2PROB", "PPH2FPR", "PPH2FDR"))

  RSID TRAININGSET      PREDICTION PPH2PROB PPH2FPR PPH2FDR
1 rs3026284 humdiv probably damaging 0.936 0.0746 0.168
2 rs3026284 humvar probably damaging 0.919 0.1330 NA
```

Detailed column descriptions can be found at `?SIFTDbColumns` and `?PolyPhenDbColumns`.

7 References

Wang K, Li M, Hakonarson H, (2010), ANNOVAR: functional annotation of genetic variants from high-throughput sequencing data. Nucleic Acids Research, Vol 38, No. 16, e164.

McLaren W, Pritchard B, RiosD, et. al., (2010), Deriving the consequences of genomic variants with the Ensembl API and SNP Effect Predictor. Bioinformatics, Vol. 26, No. 16, 2069-2070.

8 Session Information

```
R version 2.14.0 (2011-10-31)
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=C
[5] LC_MONETARY=en_US.UTF-8      LC_MESSAGES=en_US.UTF-8
[7] LC_PAPER=C                  LC_NAME=C
[9] LC_ADDRESS=C                LC_TELEPHONE=C
[11] LC_MEASUREMENT=en_US.UTF-8  LC_IDENTIFICATION=C

attached base packages:
[1] stats      graphics   grDevices utils      datasets  methods
[7] base

other attached packages:
[1] PolyPhen.Hsapiens.dbSNP131_1.0.0
[2] SIFT.Hsapiens.dbSNP132_1.0.0
[3] RSQLite_0.10.0
[4] DBI_0.2-5
[5] SNPlocs.Hsapiens.dbSNP.20090506_0.99.6
[6] BSgenome.Hsapiens.UCSC.hg18_1.3.17
[7] BSgenome_1.22.0
[8] TxDb.Hsapiens.UCSC.hg18.knownGene_2.6.2
[9] GenomicFeatures_1.6.2
[10] VariantAnnotation_1.0.5
[11] Rsamtools_1.6.1
[12] AnnotationDbi_1.16.2
[13] Biobase_2.14.0
[14] Biostrings_2.22.0
[15] GenomicRanges_1.6.3
[16] IRanges_1.12.1

loaded via a namespace (and not attached):
[1] Matrix_1.0-1        RCurl_1.7-0       XML_3.4-3
[4] biomaRt_2.10.0     bitops_1.0-4.1    grid_2.14.0
[7] lattice_0.20-0     rtracklayer_1.14.2 .snpStats_1.4.0
[10] splines_2.14.0    survival_2.36-10   tools_2.14.0
[13] zlibbioc_1.0.0
```