

# Introduction to *VariantAnnotation*

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## 1 Introduction

This vignette outlines a general workflow for annotating and filtering genetic variants using the *VariantAnnotation* package. Sample data are in VariantCall Format (VCF) and are a subset of chromosome 22 from 1000 Genomes, <ftp://ftp-trace.ncbi.nih.gov/1000genomes/ftp/release/20110521/>. VCF is a text file format that contains meta-information lines, a header line with column names, data lines with information about a position in the genome, and optional genotype information on samples for each position. A full description of the VCF format can be found on the 1000 Genomes page, <http://www.1000genomes.org/wiki/Analysis/Variant%20Call%20Format/vcf-variant-call-format-version-41>

Sample data are read in from a VCF file and variants are identified according to region such as `coding`, `intron`, `intergenic`, `spliceSite` etc. Amino acid coding changes are computed for the non-synonymous variants and SIFT and PolyPhen databases provide predictions of how severely the coding changes affect protein function. The end of the vignette covers other transformations of VCF data such as the creation of a `SnpMatrix` or a 'long form' `GRanges`.

## 2 Variant Call Format (VCF) files

### 2.1 Import complete files

Data are parsed into a VCF object with `readVcf`.

```
> library(VariantAnnotation)
> fl <- system.file("extdata", "chr22.vcf.gz", package="VariantAnnotation")
> vcf <- readVcf(fl, "hg19")
> vcf
```

```
class: VCF
dim: 10376 5
genome: hg19
exptData(1): header
fixed(4): REF ALT QUAL FILTER
info(22): LDAF AVGPOST ... VT SNPSOURCE
geno(3): GT DS GL
rownames(10376): rs7410291 rs147922003 ... rs144055359
               rs114526001
rowData values names(1): paramRangeID
colnames(5): HG00096 HG00097 HG00099 HG00100 HG00101
colData names(1): Samples
```

Extract the header information stored in the `exptData` slot

```
> hdr <- exptData(vcf)[["header"]]
> hdr
```

```
class: VCFHeader
samples(5): HG00096 HG00097 HG00099 HG00100 HG00101
meta(1): fileformat
fixed(1): ALT
info(22): LDAF AVGPOST ... VT SNPSOURCE
geno(3): GT DS GL
```

and explore it with the `fixed`, `info` and `geno` accessors. More information on this object can be found at `?VCFHeader`.

```
> fixed(hdr)
```

```
SimpleDataFrameList of length 1
names(1): ALT
```

```
> head(info(hdr), 3)
```

```
DataFrame with 3 rows and 3 columns
```

	Number	Type
	<character>	<character>
LDAF	1	Float
AVGPOST	1	Float
RSQ	1	Float

Description
<character>

LDAF                    MLE Allele Frequency Accounting for LD  
 AVGPOST Average posterior probability from MaCH/Thunder  
 RSQ                    Genotype imputation quality from MaCH/Thunder

The `GRanges` in the `rowData` slot is created from information in the `CHROM`, `POS`, and `ID` fields of the VCF file. Values in the `paramRangeID` column are meaningful when ranges have been specified in the `param` argument to `readVcf`. This is discussed further in the `Data Subsets` section.

```
> head(rowData(vcf))
```

`GRanges` with 6 ranges and 1 metadata column:

	seqnames	ranges	strand	paramRangeID
	<Rle>	<IRanges>	<Rle>	<factor>
rs7410291	22	[50300078, 50300078]	*	<NA>
rs147922003	22	[50300086, 50300086]	*	<NA>
rs114143073	22	[50300101, 50300101]	*	<NA>
rs141778433	22	[50300113, 50300113]	*	<NA>
rs182170314	22	[50300166, 50300166]	*	<NA>
rs115145310	22	[50300187, 50300187]	*	<NA>

---

seqlengths:  
 22  
 NA

The `REF`, `ALT`, `QUAL` and `FILTER` fields can be accessed together with `fixed` accessor or individually with `ref`, `alt`, `qual` and `filt` accessors.

```
> head(fixed(vcf), 3)
```

`GRanges` with 3 ranges and 5 metadata columns:

	seqnames	ranges	strand	paramRangeID
	<Rle>	<IRanges>	<Rle>	<factor>
rs7410291	22	[50300078, 50300078]	*	<NA>
rs147922003	22	[50300086, 50300086]	*	<NA>
rs114143073	22	[50300101, 50300101]	*	<NA>

  

	REF	ALT	QUAL	FILTER
	<DNAStrngSet>	<DNAStrngSetList>	<numeric>	<character>
rs7410291	A	#####	100	PASS
rs147922003	C	#####	100	PASS
rs114143073	G	#####	100	PASS

---

seqlengths:  
 22  
 NA

The `ALT` column is stored as a `DNAStrngSetList` unless the file is a structural VCF, in which case it is stored as a `CharacterList`. Extract `ALT` from the `GRanges` and determine the number of elements in the list.

```
> alternate <- alt(vcf)
> alternate
```

```

DNAStrngSetList of length 10376
[[1]] G
[[2]] T
[[3]] A
[[4]] T
[[5]] T
[[6]] A
[[7]] C
[[8]] A
[[9]] A
[[10]] C
...
<10366 more elements>

> ## number of ALT values per variant
> unique(elementLengths(alternate))

[1] 1

> head(unlist(alternate))

A DNAStrngSet instance of length 6
width seq
[1] 1 G
[2] 1 T
[3] 1 A
[4] 1 T
[5] 1 T
[6] 1 A

```

Genotype data described in the **FORMAT** field are parsed into matrices or arrays and can be accessed with the **geno** accessor. These data are not returned with the **GRanges** from **rowData** because they are unique for each sample and the data structures can be multidimensional. This is in contrast to the **fixed** and **info** data which are the same for a each variant across all samples.

Extract the header information for the genotypes.

```

> geno(hdr)

DataFrame with 3 rows and 3 columns
      Number      Type      Description
<character> <character> <character>
GT          1      String      Genotype
DS          1      Float Genotype dosage from MaCH/Thunder
GL          .      Float      Genotype Likelihoods

```

Elements of the genotype list can be accessed in the usual way.

```

> geno(vcf)

SimpleList of length 3
names(3): GT DS GL

> geno(vcf)$GT[1:3,1:5]

```

	HG000096	HG000097	HG000099	HG00100	HG00101
rs7410291	"0 0"	"0 0"	"1 0"	"0 0"	"0 0"
rs147922003	"0 0"	"0 0"	"0 0"	"0 0"	"0 0"
rs114143073	"0 0"	"0 0"	"0 0"	"0 0"	"0 0"

```
> geno(vcf)$DS[1:3,1:5]
```

	HG000096	HG000097	HG000099	HG00100	HG00101
rs7410291	0	0	1	0	0
rs147922003	0	0	0	0	0
rs114143073	0	0	0	0	0

## 2.2 Import data subsets

When working with large VCF files it may be more efficient to read in subsets of the data. Data can be subset by selecting genomic coordinates (ranges) or by selecting fields from the VCF file.

### 2.2.1 Genomic coordinates

Subset by genomic coordinates by creating a `GRanges`, `RangedData` or `RangesList`. To read in a portion of chromosome 22, we create a `GRanges` with the regions of interest.

```
> rng <- GRanges(seqnames="22",
+               ranges=IRanges(c(50301422, 50989541), c(50312106, 51001328)))
> names(rng) <- c("gene_79087", "gene_644186")
```

When ranges are specified, the VCF file must have an accompanying Tabix index file; if one does not exist it must be created. See `?indexTabix` for help creating an index.

Once the index exists a `TabixFile` instance can be created, see `?TabixFile`. This object creates a reference to the VCF and its index. Once opened, the reference remains open across calls to methods, avoiding costly index re-loading. An index file for our sample data is included in the package so the `TabixFile` can be created with,

```
> tab <- TabixFile(fl)
> tab

class: TabixFile
path: /private/tmp/RtmpKkg7Ab/Rinstafda6ba8a69a/VariantA.../chr22.vcf.gz
index: /private/tmp/RtmpKkg7Ab/Rinstafda6ba8a69a/Var.../chr22.vcf.gz.tbi
isOpen: FALSE
yieldSize: NA
```

Call `readVcf` with `TabixFile` and the ranges as the `param`. The dimension of the resulting VCF object shows 397 records overlapped with the specified ranges.

```
> vcf_rng <- readVcf(tab, "hg19", rng)
> vcf_rng
```

```
class: VCF
dim: 397 5
genome: hg19
exptData(1): header
fixed(4): REF ALT QUAL FILTER
info(22): LDAF AVGPOST ... VT SNPSOURCE
```

```

geno(3): DS GL GT
rownames(397): rs114335781 rs8135963 ... rs144055359
               rs114526001
rowData values names(1): paramRangeID
colnames(5): HG00096 HG00097 HG00099 HG00100 HG00101
colData names(1): Samples

```

The `paramRangeID` column now has meaning as it distinguishes which variant records came from which param range.

```
> head(rowData(vcf_rng), 3)
```

GRanges with 3 ranges and 1 metadata column:

	seqnames	ranges	strand	paramRangeID
	<Rle>	<IRanges>	<Rle>	<factor>
rs114335781	22	[50301422, 50301422]	*	gene_79087
rs8135963	22	[50301476, 50301476]	*	gene_79087
22:50301488	22	[50301488, 50301488]	*	gene_79087
---				
seqlengths:				
22				
NA				

### 2.2.2 VCF fields

In addition to specifying ranges, data can be subset on specific fields in the VCF file. Fields available for import are described in the header information. To view the header before reading in the data in use `ScanVcfHeader`.

```

> hdr <- scanVcfHeader(fl)
> hdr

class: VCFHeader
samples(5): HG00096 HG00097 HG00099 HG00100 HG00101
meta(1): fileformat
fixed(1): ALT
info(22): LDAF AVGPOST ... VT SNPSOURCE
geno(3): GT DS GL

```

The `info` and `geno` accessors return `DataFrames` containing descriptions of the fields, data type and number of values. A listing of all possible `info` or `geno` values is constructed by selecting the rownames of the `DataFrames`.

```

> ## INFO fields
> info_DF <- info(hdr)
> rownames(info_DF)

[1] "LDAF"      "AVGPOST"   "RSQ"       "ERATE"     "THETA"
[6] "CIEND"     "CIPOS"     "END"       "HOMLEN"    "HOMSEQ"
[11] "SVLEN"     "SVTYPE"    "AC"        "AN"        "AA"
[16] "AF"        "AMR_AF"    "ASN_AF"    "AFR_AF"    "EUR_AF"
[21] "VT"        "SNPSOURCE"

```

```
> ## FORMAT fields
> geno_DF <- geno(hdr)
> rownames(geno_DF)
```

```
[1] "GT" "DS" "GL"
```

We are interested in "LDAF" in INFO which is 'allele frequency accounting for linkage disequilibrium', and "GT" in FORMAT which is 'genotype'. Full descriptions of the elements can be seen in the header INFO and FORMAT DataFrames.

```
> info_DF[rownames(info_DF) == "LDAF", ]
```

```
DataFrame with 1 row and 3 columns
      Number      Type      Description
<character> <character> <character>
LDAF      1      Float MLE Allele Frequency Accounting for LD
```

```
> geno_DF[rownames(geno_DF) == "GT", ]
```

```
DataFrame with 1 row and 3 columns
      Number      Type Description
<character> <character> <character>
GT      1      String   Genotype
```

To subset on "LDAF" and "GT" we specify them as `character` vectors in the `info` and `geno` arguments to `ScanVcfParam`. This creates a `ScanVcfParam` object which is used as the `param` argument to `readVcf`.

```
> ## Return "ALT" from 'fixed', "LAF" from 'info' and "GT" from 'geno'
> svp <- ScanVcfParam(fixed="ALT", info="LDAF", geno="GT")
> ## Return all 'fixed' fields, "LAF" from 'info' and "GT" from 'geno'
> svp <- ScanVcfParam(info="LDAF", geno="GT")
> svp
```

```
class: ScanVcfParam
vcfWhich: 0 elements
vcfFixed: character() [All]
vcfInfo: LDAF
vcfGeno: GT
```

Note that subsetting by the VCF fields does not affect the number of ranges read in. Instead the results of the filtering are reflected in the names of the elements returned from the `info` and `geno` accessors.

```
> vcf_flds <- readVcf(fl, "hg19", svp)
> geno(vcf_flds)
```

```
SimpleList of length 1
names(1): GT
```

```
> head(info(vcf_flds), 3)
```

```
GRanges with 3 ranges and 2 metadata columns:
      seqnames      ranges strand | paramRangeID
      <Rle>      <IRanges> <Rle> | <factor>
rs7410291      22 [50300078, 50300078] * | <NA>
```

```

rs147922003      22 [50300086, 50300086]      * |      <NA>
rs114143073      22 [50300101, 50300101]      * |      <NA>
               LDAF
               <numeric>
rs7410291        0.3431
rs147922003      0.0091
rs114143073      0.0098
---
seqlengths:
22
NA

```

In the previous section we saw that a Tabix index file must exist when data are subset by genomic coordinates (i.e., ranges). This is not the case when subsetting on INFO and FORMAT elements. An index file is only needed when subsetting by ranges.

### 2.2.3 Subset on both genomic coordinates and VCF fields

To subset on both genomic coordinates and INFO and FORMAT fields the `ScanVcfParam` object must contain both. Our previous `ScanVcfParam` did not have ranges associated with it so we create a new instance with the ranges and INFO and FORMAT fields.

```

> svp_all <- ScanVcfParam(info="LDAF", geno="GT", which=rng)
> svp_all

class: ScanVcfParam
vcfWhich: 1 elements
vcfFixed: character() [All]
vcfInfo: LDAF
vcfGeno: GT

```

The subsetting here involves genomic coordinates so we need to use the Tabix index file we created.

```

> readVcf(tab, "hg19", svp_all)

class: VCF
dim: 397 5
genome: hg19
exptData(1): header
fixed(4): REF ALT QUAL FILTER
info(1): LDAF
geno(1): GT
rownames(397): rs114335781 rs8135963 ... rs144055359
               rs114526001
rowData values names(1): paramRangeID
colnames(5): HG00096 HG00097 HG00099 HG00100 HG00101
colData names(1): Samples

```

## 2.3 Adjusting chromosome names

When functions involve the comparison of ranges by overlaps. For overlap methods to work properly the chromosome names (seqlevels) must be compatible.

The VCF data chromosome names are represented by number, i.e. '22',



```
> rowdat <- rowData(vcf)
> seqlevels(rowdat)
```

```
[1] "22"
```

but the TxDb chromosome names are preceded with 'chr'.

```
> library(TxDb.Hsapiens.UCSC.hg19.knownGene)
> txdb <- TxDb.Hsapiens.UCSC.hg19.knownGene
> head(seqlevels(txdb))
```

```
[1] "chr1" "chr2" "chr3" "chr4" "chr5" "chr6"
```

Chromosome names can be modified with the `renameSeqlevels` function. Seqlevels are modified at the `GRanges` level in the `rowData` slot of the `VCF` which means all future data extractions from this `VCF` will have the new seqlevels. If the data are read in from the file again, however, the seqlevels will need to be adjusted again. See `?VCF` and `?renameSeqlevels` for examples with `VCF` and `GRanges` objects.

```
> ## rename variant seqlevels in the VCF object
> vcf <- renameSeqlevels(vcf, c("22"="chr22"))
> ## extract the rowData with modified seqlevels
> rd <- rowData(vcf)
> ## confirm seqlevels are the same
> intersect(seqlevels(rd), seqlevels(txdb))
```

```
[1] "chr22"
```

To subset a `VCF` or `GRanges` by chromosome use `keepSeqlevels`. As an example we extract transcripts for all chromosomes in `TxDb.Hsapiens.UCSC.hg19.knownGene` then keep only 'chr21' and 'chr22'. See `?VCF` and `?keepSeqlevels` for details.

```
## initially there are 93 chromosomes
> rngs <- transcripts(txdb)
> length(seqlevels(rngs))
[1] 93
## keep only chr21 and chr22
> rngs <- keepSeqlevels(rngs, c("chr21", "chr22"))
> seqlevels(rngs)
[1] "chr21" "chr22"
```

### 3 Variant location

`locateVariants` identifies where the ranges in `query` fall with respect to the annotation supplied in `subject`. Regions are specified in the `region` argument and can be one of the following constructors: `CodingVariants`, `IntronVariants`, `FiveUTRVariants`, `ThreeUTRVariants`, `IntergenicVariants`, or `SpliceSiteVariants`. Location definitions are shown in Table 1.

When the `query` is a `VCF` the variant ranges are taken from the `rowData` slot. If `query` is a `GRanges` it can have additional `elementMetadata` columns but they are ignored. As an alternative to a `TranscriptDb`, the `subject` can be a `GRangesList` of the appropriate type. `CodingVariants` would require coding regions by transcript, for `IntronVariants` introns by transcripts would be necessary, etc. See `?locateVariants` man page for details.

Identify the coding variants,

Location	Details
coding	falls <i>within</i> a coding region
fiveUTR	falls <i>within</i> a 5' untranslated region
threeUTR	falls <i>within</i> a 3' untranslated region
intron	falls <i>within</i> an intron region
intergenic	does not fall <i>within</i> a transcript associated with a gene
spliceSite	overlaps any portion of the first 2 or last 2 nucleotides of an intron
promoter	falls <i>within</i> a promoter region of a transcript

Table 1: Variant locations

```
> loc <- locateVariants(rd, txdb, CodingVariants())
> head(loc, 4)
```

GRanges with 4 ranges and 7 metadata columns:

	seqnames	ranges	strand	LOCATION	QUERYID
	<Rle>	<IRanges>	<Rle>	<factor>	<integer>
[1]	chr22	[50301422, 50301422]	*	coding	24
[2]	chr22	[50301476, 50301476]	*	coding	25
[3]	chr22	[50301488, 50301488]	*	coding	26
[4]	chr22	[50301494, 50301494]	*	coding	27

  

	TXID	CDSID	GENEID	PRECEDEID	FOLLOWID
	<integer>	<integer>	<character>	<character>	<character>
[1]	73482	217009	79087	<NA>	<NA>
[2]	73482	217009	79087	<NA>	<NA>
[3]	73482	217009	79087	<NA>	<NA>
[4]	73482	217009	79087	<NA>	<NA>

---

seqlengths:

```
chr22
NA
```

SpliceSiteVariants are those overlapping the first 2 or last 2 nucleotides of an intron.

```
> head(locateVariants(rd, txdb, SpliceSiteVariants()), 4)
```

GRanges with 4 ranges and 7 metadata columns:

	seqnames	ranges	strand	LOCATION	QUERYID
	<Rle>	<IRanges>	<Rle>	<factor>	<integer>
[1]	chr22	[50302891, 50302891]	*	spliceSite	56
[2]	chr22	[50754200, 50754202]	*	spliceSite	6618
[3]	chr22	[50960682, 50960682]	*	spliceSite	9740
[4]	chr22	[50960682, 50960682]	*	spliceSite	9740

  

	TXID	CDSID	GENEID	PRECEDEID	FOLLOWID
	<integer>	<integer>	<character>	<character>	<character>
[1]	73482	<NA>	79087	<NA>	<NA>
[2]	73514	<NA>	414918	<NA>	<NA>
[3]	72629	<NA>	29781	<NA>	<NA>
[4]	72630	<NA>	29781	<NA>	<NA>

---

seqlengths:

```
chr22
NA
```

To locate variants in all regions use the `AllVariants()` constructor,

```
> allvar <- locateVariants(rd, txdb, AllVariants())
```

The `GRanges` output of `locateVariants` includes only the ranges that fell in the specified region. Each row is a variant-transcript match which may result in multiple rows for each variant. `elementMetadata` columns returned include `LOCATION`, `QUERYID`, `TXID`, `CDSID`, and `GENEID`. In the case of `IntergenicVariants` columns for `PRECEDEID` and `FOLLOWID` are also included. The `QUERYID` column maps back to the row number in the original query.

To answer gene-centric questions data can be summarized by gene regardless of transcript.

```
> ## Did any coding variants match more than one gene?
> table(sapply(split(values(loc)[["GENEID"]], values(loc)[["QUERYID"]]),
+   function(x) length(unique(x)) > 1))
```

```
FALSE  TRUE
956    15
```

```
> ## Summarize the number of coding variants by gene ID
> idx <- sapply(split(values(loc)[["QUERYID"]], values(loc)[["GENEID"]]), unique)
> sapply(idx, length)
```

```
113730  1890  23209  23654  29781  400935  414918  415116  440836  54456
      22    15    30    87    44    15    33    11    5    82
55586   5600  56666  6300   6305  644186  79087   79174  79924  80305
      24    16    19    38    56    5    25    50    4    26
83642  83933  85378  91289  9701   9997
      55    50   147    29    68    15
```

## 4 Amino acid coding changes

`predictCoding` computes amino acid coding changes for non-synonymous variants. Only ranges in `query` that overlap with a coding region in the `subject` are considered. Reference sequences are retrieved from either a `BSgenome` or fasta file specified in `seqSource`. Variant sequences are constructed by substituting, inserting or deleting values in the `varAllele` column into the reference sequence. Amino acid codes are computed for the variant codon sequence when the length is a multiple of 3. Examples of coding situations are shown in Table 2.

The `query` argument to `predictCoding` can be a `GRanges` or `VCF`. When a `GRanges` is supplied the `varAllele` argument must be specified. In the case of a `VCF`, the alternate alleles are taken from `values(alt(<VCF>))["ALT"]` and the `varAllele` argument is not specified.

The result is a modified `query` containing only variants that fall within coding regions. Each row represents a variant-transcript match so more than one row per original variant is possible.

```
> library(BSgenome.Hsapiens.UCSC.hg19)
> coding <- predictCoding(vcf, txdb, seqSource=Hsapiens)
> coding[5:9]
```

`GRanges` with 5 ranges and 13 metadata columns:

```
      seqnames      ranges strand | paramRangeID
      <Rle>      <IRanges> <Rle> |      <factor>
```

Type	refAllele	varAllele	refCodon	varCodon	translation possible
substitution	G	T	aag	aaT	yes
substitution	G	TG	tga	tTGa	no
substitution	G	TGCG	gtc	TGCGtc	yes
insertion	“	G	cgg	Gcgg	no
insertion	“	TTG	gaa	gaTTGa	yes
deletion	A	“	atc	tc	no
deletion	GGCCTA	“	acggcctaa	aca	yes

Table 2: Amino acid coding

```

22:50301584    chr22 [50301584, 50301584]    - |    <NA>
rs114264124    chr22 [50302962, 50302962]    - |    <NA>
rs149209714    chr22 [50302995, 50302995]    - |    <NA>
22:50303554    chr22 [50303554, 50303554]    - |    <NA>
rs12167668     chr22 [50303561, 50303561]    - |    <NA>

      varAllele      CDSLOC      PROTEINLOC
<DNAStringSet> <IRanges> <CompressedIntegerList>
22:50301584      A [777, 777]      259
rs114264124      A [698, 698]      233
rs149209714      C [665, 665]      222
22:50303554      G [652, 652]      218
rs12167668      A [645, 645]      215

      QUERYID      TXID      CDSID      GENEID      CONSEQUENCE
<integer> <character> <integer> <character> <factor>
22:50301584      28      73482      217009      79087      synonymous
rs114264124      57      73482      217010      79087      nonsynonymous
rs149209714      58      73482      217010      79087      nonsynonymous
22:50303554      73      73482      217011      79087      nonsynonymous
rs12167668      74      73482      217011      79087      synonymous

      REFCODON      VARCODON      REFAA      VARAA
<DNAStringSet> <DNAStringSet> <AAStringSet> <AAStringSet>
22:50301584      CCG      CCA      P      P
rs114264124      CGG      CAG      R      Q
rs149209714      GGA      GCA      G      A
22:50303554      ATC      GTC      I      V
rs12167668      CCG      CCA      P      P
---
seqlengths:
chr22
NA

```

Using variant rs114264124 as an example, we see varAllele A has been substituted into the refCodon CCG to produce varCodon CAG. The refCodon is the sequence of codons necessary to make the variant allele substitution and therefore often includes more nucleotides than indicated in the range (i.e. the range is 50302962, 50302962, width of 1). Notice it is the second position in the refCodon that has been substituted. This position in the codon, the position of substitution, corresponds to genomic position 50302962. This genomic position maps to position 698 in coding region-based coordinates and to triplet 233 in the protein. This is a non-synonymous coding variant where the amino acid has changed from R (Arg) to Q (Gln).

When the resulting varCodon is not a multiple of 3 it cannot be translated. The consequence is considered

a frameshift and varAA will be missing.

```
> ## CONSEQUENCE is 'frameshift' where translation is not possible
> coding[values(coding)[["CONSEQUENCE"]] == "frameshift"]
```

GRanges with 1 range and 13 metadata columns:

	seqnames	ranges	strand	paramRangeID
	<Rle>	<IRanges>	<Rle>	<factor>
22:50317001	chr22	[50317001, 50317001]	+	<NA>
	varAllele	CDSLOC		PROTEINLOC
	<DNAStringSet>	<IRanges>	<CompressedIntegerList>	
22:50317001	GCACT	[808, 808]		270
	QUERYID	TXID	CDSID	GENEID CONSEQUENCE
	<integer>	<character>	<integer>	<character> <factor>
22:50317001	359	72592	214765	79174 frameshift
	REFCODON	VARCODON	REFAA	VARAA
	<DNAStringSet>	<DNAStringSet>	<AAStringSet>	<AAStringSet>
22:50317001	GCC	GCC		A

---

seqlengths:

chr22
NA

## 5 SIFT and PolyPhen Databases

From `predictCoding` we identified the amino acid coding changes for the non-synonymous variants. For this subset we can retrieve predictions of how damaging these coding changes may be. SIFT (Sorting Intolerant From Tolerant) and PolyPhen (Polymorphism Phenotyping) are methods that predict 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.

Collated predictions for specific dbSNP builds are available as downloads from the SIFT and PolyPhen web sites. These results have been packaged into *SIFT.Hsapiens.dbSNP132.db* and *PolyPhen.Hapiens.dbSNP131.db* and are designed to be searched by rsid. Variants that are in dbSNP can be searched with these database packages. When working with novel variants, SIFT and PolyPhen must be called directly. See references for home pages.

Identify the non-synonymous variants and obtain the rsids.

```
> nms <- names(coding)
> idx <- values(coding)[["CONSEQUENCE"]] == "nonsynonymous"
> nonsyn <- coding[idx]
> names(nonsyn) <- nms[idx]
> rsids <- unique(names(nonsyn)[grep("rs", names(nonsyn), fixed=TRUE)])
```

Detailed descriptions of the database columns can be found with `?SIFTDbColumns` and `?PolyPhenDbColumns`. Variants in these databases often contain more than one row per variant. The variant may have been reported by multiple sources and therefore the source will differ as well as some of the other variables.

It is important to keep in mind the pre-computed predictions in the SIFT and PolyPhen packages are based on specific gene models. SIFT is based on Ensembl and PolyPhen on UCSC Known Gene. The `TranscriptDb` we used to identify the coding snps was based on UCSC Known Gene so we will use PolyPhen for predictions. PolyPhen provides predictions using two different training datasets and has considerable

information about 3D protein structure. See `?PolyPhenDbColumns` or the PolyPhen web site listed in the references for more details.

Query the PolyPhen database,

```
> library(PolyPhen.Hsapiens.dbSNP131)
> pp <- select(PolyPhen.Hsapiens.dbSNP131, keys=rsids,
+             cols=c("TRAININGSET", "PREDICTION", "PPH2PROB"))
> head(pp[!is.na(pp$PREDICTION), ])
```

	RSID	TRAININGSET	PREDICTION	PPH2PROB
11	rs8139422	humdiv	possibly damaging	0.228
12	rs8139422	humvar	possibly damaging	0.249
13	rs74510325	humdiv	possibly damaging	0.475
14	rs74510325	humvar	possibly damaging	0.335
15	rs73891177	humdiv	benign	0.001
16	rs73891177	humvar	benign	0.005

## 6 Other operations

### 6.1 Create a SnpMatrix

The 'GT' element in the `FORMAT` field of the VCF represents the genotype. These data can be converted into a `SnpMatrix` object which can then be used with the functions offered in `snpStats` and other packages making use of the `SnpMatrix` class.

The `MatrixToSnpMatrix` function converts the genotype calls in `geno` to a `SnpMatrix`. No `dbSNP` package is used in this computation. The return value is a named list where 'genotypes' is a `SnpMatrix` and 'map' is a `DataFrame` with SNP names and alleles at each loci. The `ignore` column in 'map' indicates which variants were set to NA (missing) because they met one or more of the following criteria,

- only diploid calls are included; others are set to NA
- only single nucleotide variants are included; others are set to NA
- variants with >1 ALT allele are set to NA

See `?MatrixToSnpMatrix` for more details.

```
> calls <- geno(vcf)$GT
> a0 <- ref(vcf)
> a1 <- alt(vcf)
> res <- MatrixToSnpMatrix(calls, a0, a1)
> res
```

\$genotypes

A SnpMatrix with 5 rows and 10376 columns  
 Row names: HG00096 ... HG00101  
 Col names: rs7410291 ... rs114526001

\$map

DataFrame with 51880 rows and 4 columns

	snp.names	allele.1	allele.2	ignore
	<character>	<DNAStrngSet>	<DNAStrngSetList>	<logical>
1	rs7410291	A	#####	FALSE

2	rs147922003	C	#####	FALSE
3	rs114143073	G	#####	FALSE
4	rs141778433	C	#####	FALSE
5	rs182170314	C	#####	FALSE
6	rs115145310	G	#####	FALSE
7	rs186769856	T	#####	FALSE
8	rs77627744	G	#####	FALSE
9	rs193230365	G	#####	FALSE
...	...	...	...	...
51872	rs138542635	G	#####	FALSE
51873	rs184258531	C	#####	FALSE
51874	rs9628177	G	#####	FALSE
51875	rs9628212	G	#####	FALSE
51876	rs187302552	A	#####	FALSE
51877	rs9628178	A	#####	FALSE
51878	rs5770892	A	#####	FALSE
51879	rs144055359	G	#####	FALSE
51880	rs114526001	G	#####	FALSE

The ALT value in the 'map' **DataFrame** will be a **CharacterList** if the VCF was for structural variants or a **DNAStrngSetList** otherwise. The column is not clearly visible inside the **DataFrame** but can be extracted and inspected as follows,

```
> allele2 <- res$map[["allele.2"]]
> ## number of alternate alleles per variant
> unique(elementLengths(allele2))
```

```
[1] 1
```

```
> unlist(allele2)
```

```
A DNAStrngSet instance of length 51880
```

```
width seq
[1] 1 G
[2] 1 T
[3] 1 A
[4] 1 T
[5] 1 T
[6] 1 A
[7] 1 C
[8] 1 A
[9] 1 A
...
[51872] 1 A
[51873] 1 T
[51874] 1 A
[51875] 1 A
[51876] 1 G
[51877] 1 G
[51878] 1 G
[51879] 1 A
[51880] 1 C
```

## 6.2 Long form GRanges

The `readVcfLongForm` function reads data from a VCF file in the same manner as `readVcf` but outputs a long form `GRanges` instead of a `VCF` class. This format is driven by the fact that the alternate allele (ALT) in the VCF file often has more than one value per record. In the long form `GRanges`, the rows of the `GRanges` are replicated to match the length of the ‘unlisted’ alternate allele. This format provides access to each possible REF, ALT pair for each variant.

Input arguments and data subsetting is the same for `readVcfLongForm` as for `readVcf`. The `fixed` and `info` fields are included as `elementMetadata` columns. Currently no `geno` information is included.

`info` information was previously collected from the file header. We import ‘HOMSEQ’ and ‘ALT’.

```
> rownames(info_DF)

[1] "LDAF"      "AVGPOST"   "RSQ"       "ERATE"     "THETA"
[6] "CIEND"     "CIPOS"     "END"       "HOMLEN"    "HOMSEQ"
[11] "SVLEN"     "SVTYPE"    "AC"        "AN"        "AA"
[16] "AF"        "AMR_AF"    "ASN_AF"    "AFR_AF"    "EUR_AF"
[21] "VT"        "SNPSOURCE"
```

```
> param <- ScanVcfParam(fixed="ALT", info="HOMSEQ")
> gr <- readVcfLongForm(fl, "hg19", param)
> head(gr)
```

GRanges with 6 ranges and 5 metadata columns:

	seqnames	ranges	strand	paramRangeID	ID
	<Rle>	<IRanges>	<Rle>	<factor>	<character>
[1]	22	[50300078, 50300078]	*	<NA>	rs7410291
[2]	22	[50300086, 50300086]	*	<NA>	rs147922003
[3]	22	[50300101, 50300101]	*	<NA>	rs114143073
[4]	22	[50300113, 50300113]	*	<NA>	rs141778433
[5]	22	[50300166, 50300166]	*	<NA>	rs182170314
[6]	22	[50300187, 50300187]	*	<NA>	rs115145310

	REF	ALT	HOMSEQ
	<DNAStringSet>	<DNAStringSet>	<CompressedCharacterList>
[1]	A	G	NA
[2]	C	T	NA
[3]	G	A	NA
[4]	C	T	NA
[5]	C	T	NA
[6]	G	A	NA

```
---
seqlengths:
22
NA
```

## 6.3 Write out VCF files

A VCF file can be written out from data stored in a `VCF` class. Methods to write out from more general structures are in progress.

```
> fl <- system.file("extdata", "ex2.vcf", package="VariantAnnotation")
> out1.vcf <- tempfile()
> out2.vcf <- tempfile()
```



```
> in1 <- readVcf(fl, "hg19")
> writeVcf(in1, out1.vcf)
> in2 <- readVcf(out1.vcf, "hg19")
> writeVcf(in2, out2.vcf)
> in3 <- readVcf(out2.vcf, "hg19")
> identical(in2, in3)
```

```
[1] FALSE
```

## 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.

SIFT home page : <http://sift.bii.a-star.edu.sg/>

PolyPhen home page : <http://genetics.bwh.harvard.edu/pph2/>

## 8 Session Information

R version 2.15.3 (2013-03-01)

Platform: i386-apple-darwin9.8.0/i386 (32-bit)

locale:

```
[1] C/en_US.US-ASCII/en_US.US-ASCII/C/en_US.US-ASCII/en_US.US-ASCII
```

attached base packages:

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

other attached packages:

```
[1] snpStats_1.8.2
[2] Matrix_1.0-11
[3] lattice_0.20-13
[4] survival_2.37-4
[5] PolyPhen.Hsapiens.dbSNP131_1.0.2
[6] RSQLite_0.11.2
[7] DBI_0.2-5
[8] BSgenome.Hsapiens.UCSC.hg19_1.3.19
[9] BSgenome_1.26.1
[10] TxDb.Hsapiens.UCSC.hg19.knownGene_2.8.0
[11] GenomicFeatures_1.10.2
[12] AnnotationDbi_1.20.6
[13] Biobase_2.18.0
[14] VariantAnnotation_1.4.12
[15] Rsamtools_1.10.2
[16] Biostrings_2.26.3
```

```
[17] GenomicRanges_1.10.7  
[18] IRanges_1.16.6  
[19] BiocGenerics_0.4.0
```

loaded via a namespace (and not attached):

```
[1] RCurl_1.95-4.1      XML_3.95-0.2        biomaRt_2.14.0  
[4] bitops_1.0-5        grid_2.15.3         parallel_2.15.3  
[7] rtracklayer_1.18.2 stats4_2.15.3       tools_2.15.3  
[10] zlibbioc_1.4.0
```