

The biomaRt user's guide

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1 Introduction

In recent years a wealth of biological data has become available in public data repositories. Easy access to these valuable data resources and firm integration with data analysis is needed for comprehensive bioinformatics data analysis. The *biomaRt* package, provides an interface to a growing collection of databases implementing the BioMart software suite (<http://www.biomart.org>). The package enables retrieval of large amounts of data

in a uniform way without the need to know the underlying database schemas or write complex SQL queries. Examples of BioMart databases are Ensembl, Uniprot and HapMap. These major databases give *biomaRt* users direct access to a diverse set of data and enable a wide range of powerful online queries from R.

2 Selecting a BioMart database and dataset

Every analysis with *biomaRt* starts with selecting a BioMart database to use. A first step is to check which BioMart web services are available. The function `listMarts()` will display all available BioMart web services

```
> library("biomaRt")
> listMarts()

      biomart          version
1       ensembl ENSEMBL GENES 59 (SANGER UK)
2        .snp  ENSEMBL VARIATION 59 (SANGER UK)
3 functional_genomics ENSEMBL FUNCTIONAL GENOMICS 59 (SANGER UK)
4       vega        VEGA 38 (SANGER UK)
5 bacterial_mart_6  ENSEMBL BACTERIA 6 (EBI UK)
6   fungal_mart_6    ENSEMBL FUNGAL 6 (EBI UK)
7   metazoa_mart_6  ENSEMBL METAZOA 6 (EBI UK)
8     plant_mart_6  ENSEMBL PLANT 6 (EBI UK)
9   protist_mart_6  ENSEMBL PROTISTS 6 (EBI UK)
10       msd        MSD PROTOTYPE (EBI UK)
11      htgt HIGH THROUGHPUT GENE TARGETING AND TRAPPING (SANGER UK)
12     REACTOME      REACTOME (CSHL US)
13    wormbase215    WORMBASE 215 (CSHL US)
14      dicty      DICTYBASE (NORTHWESTERN US)
15     biomart      MGI (JACKSON LABORATORY US)
16      rgd__mart    RGD GENES (MCW US)
17    ipi_rat__mart  RGD IPI MART (MCW US)
18      SSLP__mart  RGD MICROSATELLITE MARKERS (MCW US)
19       g4public    HGNC (EBI UK)
20       pride       PRIDE (EBI UK)
21     uniprot_mart  UNIPROT (EBI UK)
22 ensembl_expressionmart_48  EURATMART (EBI UK)
23       biomartDB  PARAMECIUM GENOME (CNRS FRANCE)
24     Eurexpress Biomart  EUREXPRESS (MRC EDINBURGH UK)
25 pepseekerGOLD_mart06  PEPSEEKER (UNIVERSITY OF MANCHESTER UK)
26      Potato_01    DB_POTATO (INTERNATIONAL POTATO CENTER-CIP)
27    Sweetpotato_01  DB_SWEETPOTATO (INTERNATIONAL POTATO CENTER-CIP)
28 phytozome_mart      PHYTOZOME (JGI/CIG US)
29      cyanobase_1  CYANOBASE 1 (KAZUSA JAPAN)
30     HapMap_rel27    HAPMAP 27 (NCBI US)
31      CosmicMart    COSMIC (SANGER UK)
32     cildb_all_v2    CILDB INPARANOID AND FILTERED BEST HIT (CNRS FRANCE)
33     cildb_inp_v2    CILDB INPARANOID (CNRS FRANCE)
34 GRAMENE_MARKER_30  GRAMENE 30 MARKERS (CSHL/CORNELL US)
35 GRAMENE_MAP_30    GRAMENE 30 MAPPINGS (CSHL/CORNELL US)
```

```

36          QTL_MART           GRAMENE 30 QTL DB (CSHL/CORNELL US)
37          genes             INTOGEN GENES
38          oncomodules       INTOGEN ONCOMODULES
39          gmap_japonica     RICE-MAP JAPONICA (PEKING UNIVESITY CHINA)
40 europhenomeannotations   EUROPHENOME
41          emma_biomart      THE EUROPEAN MOUSE MUTANT ARCHIVE (EMMA)
42          ikmc              IKMC GENES AND PRODUCTS (I-DCC)
43          gmap_indica       RICE-MAP INDICA (PEKING UNIVERSITY CHINA)
44          Ensembl156        PANCREATIC EXPRESSION DATABASE (INSTITUTE OF CANCER UK)

```

Note: if the function `useMart` runs into proxy problems you should set your proxy first before calling any biomaRt functions. You can do this using the `Sys.getenv` command:

```
Sys.getenv("http\_proxy" = "http://my.proxy.org:9999")
```

The `useMart` function can now be used to connect to a specified BioMart database, this must be a valid name given by `listMarts`. In the next example we choose to query the Ensembl BioMart database.

```
> ensembl = useMart("ensembl")
```

BioMart databases can contain several datasets, for Ensembl every species is a different dataset. In a next step we look at which datasets are available in the selected BioMart by using the function `listDatasets`.

```
> listDatasets(ensembl)
```

	dataset	description	version
1	oanatinus_gene_ensembl	Ornithorhynchus anatinus genes (OANA5)	OANA5
2	tguttata_gene_ensembl	Taeniopygia guttata genes (taeGut3.2.4)	taeGut3.2.4
3	cporcellus_gene_ensembl	Cavia porcellus genes (cavPor3)	cavPor3
4	gaculeatus_gene_ensembl	Gasterosteus aculeatus genes (BROADS1)	BROADS1
5	lafricana_gene_ensembl	Loxodonta africana genes (loxAfr3)	loxAfr3
6	mlucifugus_gene_ensembl	Myotis lucifugus genes (myoLuc1)	myoLuc1
7	hsapiens_gene_ensembl	Homo sapiens genes (GRCh37)	GRCh37
8	choffmanni_gene_ensembl	Choloepus hoffmanni genes (choHof1)	choHof1
9	csavignyi_gene_ensembl	Ciona savignyi genes (CSAV2.0)	CSAV2.0
10	fcatus_gene_ensembl	Felis catus genes (CAT)	CAT
11	rnorvegicus_gene_ensembl	Rattus norvegicus genes (RGSC3.4)	RGSC3.4
12	ggallus_gene_ensembl	Gallus gallus genes (WASHUC2)	WASHUC2
13	tbelangeri_gene_ensembl	Tupaia belangeri genes (tupBel1)	tupBel1
14	xtropicalis_gene_ensembl	Xenopus tropicalis genes (JGI4.1)	JGI4.1
15	ecaballus_gene_ensembl	Equus caballus genes (EquCab2)	EquCab2
16	cjacchus_gene_ensembl	Callithrix jacchus genes (calJac3)	calJac3
17	drerio_gene_ensembl	Danio rerio genes (Zv8)	Zv8
18	stridemclineatus_gene_ensembl	Spermophilus tridemclineatus genes (speTri1)	speTri1
19	tnigroviridis_gene_ensembl	Tetraodon nigroviridis genes (TETRAODON8.0)	TETRAODON8.0
20	ttruncatus_gene_ensembl	Tursiops truncatus genes (turTru1)	turTru1
21	scerevisiae_gene_ensembl	Saccharomyces cerevisiae genes (SGD1.01)	SGD1.01
22	celegans_gene_ensembl	Caenorhabditis elegans genes (WS210)	WS210

23	mmulatta_gene_ensembl	Macaca mulatta genes (MMUL_1.0)	MMUL_1.0
24	pvampyrus_gene_ensembl	Pteropus vampyrus genes (pteVam1)	pteVam1
25	mdomestica_gene_ensembl	Monodelphis domestica genes (monDom5)	monDom5
26	vpacos_gene_ensembl	Vicugna pacos genes (vicPac1)	vicPac1
27	acarolinensis_gene_ensembl	Anolis carolinensis genes (AnoCar1.0)	AnoCar1.0
28	tsyrichta_gene_ensembl	Tarsius syrichta genes (tarSyr1)	tarSyr1
29	ogarnettii_gene_ensembl	Otolemur garnettii genes (otoGar1)	otoGar1
30	trubripes_gene_ensembl	Takifugu rubripes genes (FUGU4.0)	FUGU4.0
31	dmelanogaster_gene_ensembl	Drosophila melanogaster genes (BDGP5.13)	BDGP5.13
32	eeuropaeus_gene_ensembl	Erinaceus europaeus genes (eriEur1)	eriEur1
33	mmurinus_gene_ensembl	Microcebus murinus genes (micMur1)	micMur1
34	olatipes_gene_ensembl	Oryzias latipes genes (HdrR)	HdrR
35	etelfairi_gene_ensembl	Echinops telfairi genes (TENREC)	TENREC
36	cintestinalis_gene_ensembl	Ciona intestinalis genes (JGI2)	JGI2
37	ptroglodytes_gene_ensembl	Pan troglodytes genes (CHIMP2.1)	CHIMP2.1
38	oprinceps_gene_ensembl	Ochotona princeps genes (OchPri2.0)	OchPri2.0
39	ggorilla_gene_ensembl	Gorilla gorilla genes (gorGor3)	gorGor3
40	dordii_gene_ensembl	Dipodomys ordii genes (dipOrd1)	dipOrd1
41	ppygmaeus_gene_ensembl	Pongo pygmaeus abelii genes (PPYG2)	PPYG2
42	sscrofa_gene_ensembl	Sus scrofa genes (SScrofa9)	SScrofa9
43	mmusculus_gene_ensembl	Mus musculus genes (NCBIM37)	NCBIM37
44	ocuniculus_gene_ensembl	Oryctolagus cuniculus genes (oryCun2.0)	oryCun2.0
45	mgallopavo_gene_ensembl	Meleagris gallopavo genes (UMD2)	UMD2
46	saraneus_gene_ensembl	Sorex araneus genes (sorAra1)	sorAra1
47	dnovemcinctus_gene_ensembl	Dasypus novemcinctus genes (dasNov2)	dasNov2
48	pcapensis_gene_ensembl	Procavia capensis genes (proCap1)	proCap1
49	btaurus_gene_ensembl	Bos taurus genes (Btau_4.0)	Btau_4.0
50	meugenii_gene_ensembl	Macropus eugenii genes (Meug_1.0)	Meug_1.0
51	cfamiliaris_gene_ensembl	Canis familiaris genes (CanFam_2.0)	CanFam_2.0

To select a dataset we can update the `Mart` object using the function `useDataset`. In the example below we choose to use the `hsapiens` dataset.

```
ensembl = useDataset("hsapiens_gene_ensembl",mart=ensembl)
```

Or alternatively if the dataset one wants to use is known in advance, we can select a BioMart database and dataset in one step by:

```
> ensembl = useMart("ensembl", dataset = "hsapiens_gene_ensembl")
```

3 How to build a biomaRt query

The `getBM` function has three arguments that need to be introduced: filters, attributes and values. *Filters* define a restriction on the query. For example you want to restrict the output to all genes located on the human X chromosome then the filter `chromosome_name` can be used with value 'X'. The `listFilters` function shows you all available filters in the selected dataset.

```

> filters = listFilters(ensembl)
> filters[1:5, ]

      name      description
1 chromosome_name Chromosome name
2          start Gene Start (bp)
3          end   Gene End (bp)
4    band_start     Band Start
5    band_end       Band End

```

Attributes define the values we are interested in to retrieve. For example we want to retrieve the gene symbols or chromosomal coordinates. The `listAttributes` function displays all available attributes in the selected dataset.

```

> attributes = listAttributes(ensembl)
> attributes[1:5, ]

      name      description
1 ensemble_gene_id      Ensembl Gene ID
2 ensemble_transcript_id Ensembl Transcript ID
3 ensemble_peptide_id   Ensembl Protein ID
4 canonical_transcript_stable_id Canonical transcript stable ID(s)
5 description           Description

```

The `getBM` function is the main query function in `biomaRt`. It has four main arguments:

- `attributes`: is a vector of attributes that one wants to retrieve (= the output of the query).
- `filters`: is a vector of filters that one wil use as input to the query.
- `values`: a vector of values for the filters. In case multple filters are in use, the `values` argument requires a list of values where each position in the list corresponds to the position of the filters in the `filters` argument (see examples below).
- `mart`: is and object of class `Mart`, which is created by the `useMart` function.

Note: for some frequently used queries to Ensembl, wrapper functions are available: `getGene` and `getSequence`. These functions call the `getBM` function with hard coded filter and attribute names.

Now that we selected a BioMart database and dataset, and know about attributes, filters, and the values for filters; we can build a biomaRt query. Let's make an easy query for the following problem: We have a list of Affymetrix identifiers from the u133plus2 platform and we want to retrieve the corresponding EntrezGene identifiers using the Ensembl mappings.

The u133plus2 platform will be the filter for this query and as values for this filter we use our list of Affymetrix identifiers. As output (attributes) for the query we want to retrieve the EntrezGene and u133plus2 identifiers so we get a mapping of these two identifiers as a result. The exact names that we will have to use to specify the attributes and filters can be retrieved with the `listAttributes` and `listFilters` function respectively. Let's now run the query:

```
> affyids = c("202763_at", "209310_s_at", "207500_at")
> getBM(attributes = c("affy_hg_u133_plus_2", "entrezgene"), filters = "affy_hg_u133_plus_2",
+       values = affyids, mart = ensembl)

  affy_hg_u133_plus_2 entrezgene
1      209310_s_at      837
2      209310_s_at      NA
3      207500_at        NA
4      207500_at        838
5      202763_at        836
6      202763_at        NA
```

4 Examples of biomaRt queries

In the sections below a variety of example queries are described. Every example is written as a task, and we have to come up with a biomaRt solution to the problem.

4.1 Task 1: Annotate a set of Affymetrix identifiers with HUGO symbol and chromosomal locations of corresponding genes

We have a list of Affymetrix hgu133plus2 identifiers and we would like to retrieve the HUGO gene symbols, chromosome names, start and end positions and the bands of the corresponding genes. The `listAttributes` and the `listFilters` functions give us an overview of the available attributes and filters and we look in those lists to find the corresponding attribute and filter names we need. For this query we'll need the following attributes: hgnc_symbol, chromosome_name, start_position, end_position, band

and affy_hg_u133_plus_2 (as we want these in the output to provide a mapping with our original Affymetrix input identifiers. There is one filter in this query which is the affy_hg_u133_plus_2 filter as we use a list of Affymetrix identifiers as input. Putting this all together in the `getBM` and performing the query gives:

```
> affyids = c("202763_at", "209310_s_at", "207500_at")
> getBM(attributes = c("affy_hg_u133_plus_2", "hgnc_symbol", "chromosome_name", "start_position",
+   "end_position", "band"), filters = "affy_hg_u133_plus_2", values = affyids, mart = ensembl)

affy_hg_u133_plus_2 hgnc_symbol chromosome_name start_position end_position band
1      209310_s_at      CASP4                  11     104813594  104840163 q22.3
2      207500_at       CASP5                  11     104864962  104893895 q22.3
3      202763_at       CASP3                   4     185548850  185570629 q35.1
```

4.2 Task 2: Annotate a set of EntrezGene identifiers with GO annotation

In this task we start out with a list of EntrezGene identifiers and we want to retrieve GO identifiers related to biological processes that are associated with these entrezgene identifiers. Again we look at the output of `listAttributes` and `listFilters` to find the filter and attributes we need. Then we construct the following query:

```
> entrez = c("673", "837")
> getBM(attributes = c("entrezgene", "go_biological_process_id"), filters = "entrezgene", values = entrez,
+   mart = ensembl)

entrezgene go_biological_process_id
1      673          GO:0000165
2      673          GO:0006916
3      673          GO:0051591
4      673          GO:0043434
5      673          GO:0007264
6      673          GO:0043524
7      673          GO:0070374
8      673          GO:0006468
9      673          GO:0051291
10     673          GO:0009887
11     673          GO:0023034
12     673          GO:0007165
13     837          GO:0006508
14     837          GO:0006917
15     837          GO:0006915
16     837          GO:0042981
```

4.3 Task 3: Retrieve all HUGO gene symbols of genes that are located on chromosomes 1,2 or Y , and are associated with one the following GO terms: "GO:0051330","GO:0000080","GO:0000114","GO:0000082" (here we'll use more than one filter)

The `getBM` function enables you to use more than one filter. In this case the filter argument should be a vector with the filter names. The values should be a list, where the first element of the list corresponds to the first filter and the second list element to the second filter and so on. The elements of this list are vectors containing the possible values for the corresponding filters.

```
go=c("GO:0051330", "GO:0000080", "GO:0000114")
chrom=c(1,2, "Y")
getBM(attributes= "hgnc_symbol",
      filters=c("go", "chromosome_name"),
      values=list(go,chrom), mart=ensembl)

hgnc_symbol
1    PPP1CB
2    SPDYA
3    ACVR1
4    CUL3
5    RCC1
6    CDC7
7    RHOU
```

4.4 Task 4: Annotate set of identifiers with INTERPRO protein domain identifiers

In this example we want to annotate the following two RefSeq identifiers: NM_005359 and NM_000546 with INTERPRO protein domain identifiers and a description of the protein domains.

```
> refseqids = c("NM_005359", "NM_000546")
> ipro = getBM(attributes = c("refseq_dna", "interpro", "interpro_description"), filter
+   values = refseqids, mart = ensembl)

ipro
  refseq_dna  interpro          interpro_description
  1 NM_000546 IPR002117          p53 tumor antigen
  2 NM_000546 IPR010991          p53, tetramerisation
  3 NM_000546 IPR011615          p53, DNA-binding
  4 NM_000546 IPR013872 p53 transactivation domain (TAD)
  5 NM_000546 IPR000694          Proline-rich region
  6 NM_005359 IPR001132          MAD homology 2, Dwarfin-type
  7 NM_005359 IPR003619          MAD homology 1, Dwarfin-type
  8 NM_005359 IPR013019          MAD homology, MH1
```

4.5 Task 5: Select all Affymetrix identifiers on the hgu133plus2 chip and Ensembl gene identifiers for genes located on chromosome 16 between basepair 1100000 and 1250000.

In this example we will again use multiple filters: chromosome_name, start, and end as we filter on these three conditions. Note that when a chromosome name, a start position and an end position are jointly used as filters, the BioMart webservice interprets this as return everything from the given chromosome between the given start and end positions.

```
> getBM(c("affy_hg_u133_plus_2", "ensembl_gene_id"), filters = c("chromosome_name", "start",
+   "end"), values = list(16, 1100000, 1250000), mart = ensembl)

affy_hg_u133_plus_2 ensembl_gene_id
1           214555_at ENSG00000162009
2           ENSG00000184471
3           205845_at ENSG00000196557
4           ENSG00000181791
```

4.6 Task 6: Retrieve all entrezgene identifiers and HUGO gene symbols of genes which have a "MAP kinase activity" GO term associated with it.

The GO identifier for MAP kinase activity is GO:0004707. In our query we will use go as filter and entrezgene and hgnc_symbol as attributes. Here's the query:

```
> getBM(c("entrezgene", "hgnc_symbol"), filters = "go", values = "GO:0004707", mart = ensembl)

entrezgene hgnc_symbol
1      5596    MAPK4
2      5597    MAPK6
3       NA    MAPK7
4      5598    MAPK7
5     225689   MAPK15
6       NA   MAPK15
7      5595    MAPK3
8       NA    MAPK3
9     51701    NLK
10      NA    NLK
11     5602   MAPK10
12      NA   MAPK10
13     5599    MAPK8
14      NA    MAPK8
15     5594    MAPK1
16      NA    MAPK1
17     1432   MAPK14
18      NA   MAPK14
19     5603   MAPK13
20      NA   MAPK13
21     6300   MAPK12
22     5600   MAPK11
23     5601    MAPK9
```

4.7 Task 7: Given a set of EntrezGene identifiers, retrieve 100bp upstream promoter sequences

All sequence related queries to Ensembl are available through the `getSequence` wrapper function. `getBM` can also be used directly to retrieve sequences but this can get complicated so using `getSequence` is recommended. Sequences can be retrieved using the `getSequence` function either starting from chromosomal coordinates or identifiers. The chromosome name can be specified using the `chromosome` argument. The `start` and `end` arguments are used to specify `start` and `end` positions on the chromosome. The type of sequence returned can be specified by the `seqType` argument which takes the following values: 'cdna';'peptide' for protein sequences; '3utr' for 3' UTR sequences; '5utr' for 5' UTR sequences; 'gene_exon' for exon sequences only; 'transcript_exon' for transcript specific exonic sequences only; 'transcript_exon_intron' gives the full unspliced transcript, that is exons + introns; 'gene_exon_intron' gives the exons + introns of a gene; 'coding' gives the coding sequence only; 'coding_transcript_flank' gives the flanking region of the transcript including the UTRs, this must be accompanied with a given value for the upstream or downstream attribute; 'coding_gene_flank' gives the flanking region of the gene including the UTRs, this must be accompanied with a given value for the upstream or downstream attribute; 'transcript_flank' gives the flanking region of the transcript excluding the UTRs, this must be accompanied with a given value for the upstream or downstream attribute; 'gene_flank' gives the flanking region of the gene excluding the UTRs, this must be accompanied with a given value for the upstream or downstream attribute.

In MySQL mode the `getSequence` function is more limited and the sequence that is returned is the 5' to 3'+ strand of the genomic sequence, given a chromosome, as start and an end position.

Task 4 requires us to retrieve 100bp upstream promoter sequences from a set of EntrezGene identifiers. The type argument in `getSequence` can be thought of as the filter in this query and uses the same input names given by `listFilters`. In our query we use `entrezgene` for the type argument. Next we have to specify which type of sequences we want to retrieve, here we are interested in the sequences of the promoter region, starting right next to the coding start of the gene. Setting the `seqType` to `coding_gene_flank` will give us what we need. The `upstream` argument is used to specify how many bp of upstream sequence we want to retrieve, here we'll retrieve a rather short sequence of 100bp. Putting this all together in `getSequence` gives:

```
> entrez = c("673", "7157", "837")
> getSequence(id = entrez, type = "entrezgene", seqType = "coding_gene_flank", upstream = 100,
+   mart = ensembl)
```

4.8 Task 8: Retrieve all 5' UTR sequences of all genes that are located on chromosome 3 between the positions 185514033 and 185535839

As described in the previous task `getSequence` can also use chromosomal coordinates to retrieve sequences of all genes that lie in the given region. We also have to specify which type of identifier we want to retrieve together with the sequences, here we choose for `entrezgene` identifiers.

```
> utr5 = getSequence(chromosome = 3, start = 185514033, end = 185535839, type = "entrezgene",
+   seqType = "5utr", mart = ensembl)
> utr5
```

V1	V2
.....GAAGCGGTGGC	1981

4.9 Task 9: Retrieve protein sequences for a given list of EntrezGene identifiers

In this task the `type` argument specifies which type of identifiers we are using. To get an overview of other valid identifier types we refer to the `listFilters` function.

```
> protein = getSequence(id = c(100, 5728), type = "entrezgene", seqType = "peptide", mart = ensembl)
> protein
```

peptide	entrezgene
MAQTPAFDKPKVEL ...	100
MTAIKEIVSRNKRR ...	5728

4.10 Task 10: Retrieve known SNPs located on the human chromosome 8 between positions 148350 and 148612

For this example we'll first have to connect to a different BioMart database, namely `snp`.

```
> snpmart = useMart("snp", dataset = "hsapiens_snp")
```

The `listAttributes` and `listFilters` functions give us an overview of the available attributes and filters. From these we need: `refsnp_id`, `allele`, `chrom_start` and `chrom_strand` as attributes; and as filters we'll use:

`chrom_start`, `chrom_end` and `chr_name`. Note that when a chromosome name, a start position and an end position are jointly used as filters, the BioMart webservice interprets this as return everything from the given chromosome between the given start and end positions. Putting our selected attributes and filters into `getBM` gives:

```
> getBM(c("refsnp_id", "allele", "chrom_start", "chrom_strand"), filters = c("chr_name", "chrom_start",
+   "chrom_end"), values = list(8, 148350, 148612), mart = snpmart)

  refsnp_id allele chrom_start chrom_strand
1  rs1134195   G/T     148394      -1
2  rs4046274   C/A     148394       1
3  rs4046275   A/G     148411       1
4  rs13291     C/T     148462       1
5  rs1134192   G/A     148462      -1
6  rs4046276   C/T     148462       1
7  rs12019378  T/G     148471       1
8  rs1134191   C/T     148499      -1
9  rs4046277   G/A     148499       1
10 rs11136408  G/A     148525       1
11 rs1134190   C/T     148533      -1
12 rs4046278   G/A     148533       1
13 rs1134189   G/A     148535      -1
14 rs3965587   C/T     148535       1
15 rs1134187   G/A     148539      -1
16 rs1134186   T/C     148569       1
17 rs4378731   G/A     148601       1
```

4.11 Task 11: Given the human gene TP53, retrieve the human chromosomal location of this gene and also retrieve the chromosomal location and RefSeq id of it's homolog in mouse.

The `getLDS` (Get Linked Dataset) function provides functionality to link 2 BioMart datasets which each other and construct a query over the two datasets. In Ensembl, linking two datasets translates to retrieving homology data across species. The usage of `getLDS` is very similar to `getBM`. The linked dataset is provided by a separate `Mart` object and one has to specify filters and attributes for the linked dataset. Filters can either be applied to both datasets or to one of the datasets. Use the `listFilters` and `listAttributes` functions on both `Mart` objects to find the filters and attributes for each dataset (species in Ensembl). The attributes and filters of the linked dataset can be specified with the `attributesL` and `filtersL` arguments. Entering all this information into `getLDS` gives:

```
human = useMart("ensembl", dataset = "hsapiens_gene_ensembl")
mouse = useMart("ensembl", dataset = "mmusculus_gene_ensembl")
getLDS(attributes = c("hgnc_symbol", "chromosome_name", "start_position"),
       filters = "hgnc_symbol", values = "TP53", mart = human,
       attributesL = c("refseq_dna", "chromosome_name", "start_position"), martL = mouse)
```

```

V1 V2      V3      V4 V5      V6
1 TP53 17 7512464 NM_011640 11 69396600

```

5 Using archived versions of Ensembl

It is possible to query archived versions of Ensembl through *biomaRt*. There are currently two ways to access archived versions.

5.1 Using the archive=TRUE

First we list the available Ensembl archives by using the `listMarts` function and setting the archive attribute to TRUE. Note that not all archives are available this way and it seems that recently this only gives access to few archives if you don't see the version of the archive you need please look at the 2nd way to access archives.

```

> listMarts(archive = TRUE)

      biomart          version
1     ensembl_mart_51    Ensembl 51
2       snp_mart_51        SNP 51
3       vega_mart_51       Vega 32
4     ensembl_mart_50    Ensembl 50
5       snp_mart_50        SNP 50
6       vega_mart_50       Vega 32
7     ensembl_mart_49  ENSEMBL GENES 49 (SANGER)
8 genomic_features_mart_49   Genomic Features
9       snp_mart_49        SNP
10      vega_mart_49       Vega
11     ensembl_mart_48  ENSEMBL GENES 48 (SANGER)
12 genomic_features_mart_48   Genomic Features
13       snp_mart_48        SNP
14      vega_mart_48       Vega
15     ensembl_mart_47  ENSEMBL GENES 47 (SANGER)
16 genomic_features_mart_47   Genomic Features
17       snp_mart_47        SNP
18      vega_mart_47       Vega
19     compara_mart_homology_47  Compara homology
20 compara_mart_multiple_ga_47 Compara multiple alignments
21 compara_mart_pairwise_ga_47 Compara pairwise alignments
22     ensembl_mart_46  ENSEMBL GENES 46 (SANGER)
23 genomic_features_mart_46   Genomic Features
24       snp_mart_46        SNP
25      vega_mart_46       Vega
26     compara_mart_homology_46  Compara homology
27 compara_mart_multiple_ga_46 Compara multiple alignments
28 compara_mart_pairwise_ga_46 Compara pairwise alignments
29     ensembl_mart_45  ENSEMBL GENES 45 (SANGER)
30       snp_mart_45        SNP
31      vega_mart_45       Vega
32     compara_mart_homology_45  Compara homology
33 compara_mart_multiple_ga_45 Compara multiple alignments

```

```

34 compara_mart_pairwise_ga_45 Compara pairwise alignments
35         ensembl_mart_44    ENSEMBL GENES 44 (SANGER)
36            .snp_mart_44          SNP
37                 vega_mart_44        Vega
38     compara_mart_homology_44           Compara homology
39 compara_mart_pairwise_ga_44 Compara pairwise alignments
40         ensembl_mart_43    ENSEMBL GENES 43 (SANGER)
41            .snp_mart_43          SNP
42                 vega_mart_43        Vega
43     compara_mart_homology_43           Compara homology
44 compara_mart_pairwise_ga_43 Compara pairwise alignments

```

Next we select the archive we want to use using the `useMart` function, again setting the archive attribute to TRUE and giving the full name of the BioMart e.g. `ensembl_mart_46`.

```
> ensembl = useMart("ensembl_mart_46", dataset = "hsapiens_gene_ensembl", archive = T)
```

If you don't know the dataset you want to use could first connect to the BioMart using `useMart` and then use the `listDatasets` function on this object. After you selected the BioMart database and dataset, queries can be performed in the same way as when using the current BioMart versions.

5.2 Accessing archives through specifying the archive host

Use the <http://www.ensembl.org> website and go down the bottom of the page. Click on 'view in Archive' and select the archive you need. Copy the url and use that url as shown below to connect to the specified BioMart database. The example below shows how to query Ensembl 54.

```
> listMarts(host = "may2009.archive.ensembl.org")
> ensembl54 = useMart(host = "may2009.archive.ensembl.org", biomart = "ENSEMBL_MART_ENSEMBL")
> ensembl54 = useMart(host = "may2009.archive.ensembl.org", biomart = "ENSEMBL_MART_ENSEMBL",
+   dataset = "hsapiens_gene_ensembl")
```

6 Using a BioMart other than Ensembl

To demonstrate the use of the `biomaRt` package with non-Ensembl databases the next query is performed using the Wormbase BioMart (WormMart). We connect to Wormbase, select the gene dataset to use and have a look at the available attributes and filters. Then we use a list of gene names as filter and retrieve associated RNAi identifiers together with a description of the RNAi phenotype.

```
> wormbase = useMart("wormbase_current", dataset = "wormbase_gene")
> listFilters(wormbase)
> listAttributes(wormbase)
> getBM(attributes = c("name", "rnai", "rnai_phenotype", "phenotype_desc"), filters = "gene_name",
+   values = c("unc-26", "his-33"), mart = wormbase)
```

	name rnai	rnai_phenotype	phenotype_desc
1	his-33 WBRNAi00000104	Emb Nmo	embryonic lethal Nuclear morphology alteration in early embryo
2	his-33 WBRNAi00012233	WT	wild type morphology
3	his-33 WBRNAi00024356	Ste	sterile
4	his-33 WBRNAi00025036	Emb	embryonic lethal
5	his-33 WBRNAi00025128	Emb	embryonic lethal
6	his-33 WBRNAi00025393	Emb	embryonic lethal
7	his-33 WBRNAi00025515	Emb Lva Unc	embryonic lethal larval arrest uncoordinated
8	his-33 WBRNAi00025632	Gro Ste	slow growth sterile
9	his-33 WBRNAi00025686	Gro Ste	slow growth sterile
10	his-33 WBRNAi00025785	Gro Ste	slow growth sterile
11	his-33 WBRNAi00026259	Emb Gro Unc	embryonic lethal slow growth uncoordinated
12	his-33 WBRNAi00026375	Emb	embryonic lethal
13	his-33 WBRNAi00026376	Emb	embryonic lethal
14	his-33 WBRNAi00027053	Emb Unc	embryonic lethal uncoordinated
15	his-33 WBRNAi00030041	WT	wild type morphology
16	his-33 WBRNAi00031078	Emb	embryonic lethal
17	his-33 WBRNAi00032317	Emb	embryonic lethal
18	his-33 WBRNAi00032894	Emb	embryonic lethal
19	his-33 WBRNAi00033648	Emb	embryonic lethal
20	his-33 WBRNAi00035430	Emb	embryonic lethal
21	his-33 WBRNAi00035860	Egl Emb	egg laying defect embryonic lethal
22	his-33 WBRNAi00048335	Emb Sister Chromatid Separation abnormal (Cross-eyed)	embryonic lethal
23	his-33 WBRNAi00049266	Emb Sister Chromatid Separation abnormal (Cross-eyed)	embryonic lethal
24	his-33 WBRNAi00053026	Emb Sister Chromatid Separation abnormal (Cross-eyed)	embryonic lethal
25	unc-26 WBRNAi00021278	WT	wild type morphology
26	unc-26 WBRNAi00026915	WT	wild type morphology
27	unc-26 WBRNAi00026916	WT	wild type morphology
28	unc-26 WBRNAi00027544	Unc	uncoordinated
29	unc-26 WBRNAi00049565	WT	wild type morphology
30	unc-26 WBRNAi00049566	WT	wild type morphology

7 biomaRt helper functions

This section describes a set of biomaRt helper functions that can be used to export FASTA format sequences, retrieve values for certain filters and exploring the available filters and attributes in a more systematic manner.

7.1 exportFASTA

The data.frames obtained by the getSequence function can be exported to FASTA files using the `exportFASTA` function. One has to specify the data.frame to export and the filename using the file argument.

7.2 Finding out more information on filters

7.2.1 filterType

Boolean filters need a value TRUE or FALSE in biomaRt. Setting the value TRUE will include all information that fulfill the filter requirement. Setting FALSE will exclude the information that fulfills the filter requirement and will return all values that don't fulfill the filter. For most of the filters, their name indicates if the type is a boolean or not and they will usually start with "with". However this is not a rule and to make sure you got the type right you can use the function `filterType` to investigate the type of the filter you want to use.

```
> filterType("with_affy_hg_u133_plus_2", ensembl)
[1] "boolean_list"
```

7.2.2 filterOptions

Some filters have a limited set of values that can be given to them. To know which values these are one can use the `filterOptions` function to retrieve the predetermined values of the respective filter.

```
> filterOptions("biotype", ensembl)
[1] "[IG_C_gene,IG_D_gene,IG_J_gene,IG_J_pseudogene,IG_pseudogene,IG_V_gene,IG_V_pseudogene,...]
```

If there are no predetermined values e.g. for the entrezgene filter, then `filterOptions` will return the type of filter it is. And most of the times the filter name or its description will suggest what values one can use for the respective filter (e.g. entrezgene filter will work with entrezgene identifiers as values)

7.3 Attribute Pages

For large BioMart databases such as Ensembl, the number of attributes displayed by the `listAttributes` function can be very large. In BioMart databases, attributes are put together in pages, such as sequences, features, homologs for Ensembl. An overview of the attributes pages present in the respective BioMart dataset can be obtained with the `attributePages` function.

```
> pages = attributePages(ensembl)
> pages
```

```
[1] "feature_page"      "structure"          "transcript_event" "homologs"        "snp"
```

To show us a smaller list of attributes which belong to a specific page, we can now specify this in the `listAttributes` function as follows:

```
> listAttributes(ensembl, page = "feature_page")
```

1	name	des
2	ensembl_gene_id	Ensembl
3	ensembl_transcript_id	Ensembl Trans
4	ensembl_peptide_id	Ensembl Pr
5	canonical_transcript_stable_id	Canonical transcript stab
6	description	Des
7	chromosome_name	Chromos
8	start_position	Gene St
9	end_position	Gene
10	strand	
11	band	
12	transcript_start	Transcript St
13	transcript_end	Transcript E
14	external_gene_id	Associated G
15	external_transcript_id	Associated Transcri
16	external_gene_db	Associated
17	transcript_db_name	Associated Trans
18	transcript_count	Transcri
19	percentage_gc_content	% GC
20	gene_biotype	Gene
21	transcript_biotype	Transcript
22	source	
23	status	Status
24	transcript_status	Status (tra
25	go_biological_process_id	GO Term Access
26	name_1006	GO Term N
27	definition_1006	GO Term Definit
28	go_biological_process_linkage_type	GO Term Evidence C
29	go_cellular_component_id	GO Term Access
30	go_cellular_component__dm_name_1006	GO Term N
31	go_cellular_component__dm_definition_1006	GO Term Definit
32	go_cellular_component_linkage_type	GO Term Evidence C
33	go_molecular_function_id	GO Term A
34	go_molecular_function__dm_name_1006	GO Term N
35	go_molecular_function__dm_definition_1006	GO Term Definit
36	go_molecular_function_linkage_type	GO Term Evidence C
37	goslim_goa_accession	GOSlim GOA Acce
38	goslim_goa_description	GOSlim GOA Des
	ucsc	

```

39                      pdb
40          clone_based_ensembl_gene_name
41          clone_based_ensembl_transcript_name
42              clone_based_vega_gene_name
43              clone_based_vega_transcript_name
44                  ccds
45                  embl
46 ox_ens_lrg_transcript__dm_dbprimary_acc_1074
47                      entrezgene
48                          ottt
49                          ottg
50              shares_cds_with_enst
51              shares_cds_with_ottt
52      shares_cds_and_utr_with_ottt
53          hgnc_id
54          hgnc_symbol
55          hgnc_automatic_gene_name
56          hgnc_curated_gene_name
57          hgnc_automatic_transcript_name
58          hgnc_curated_transcript_name
59          hgnc_mb001
60          ipi
61          merops
62          mim_morbid_accession
63          mim_morbid_description
64          mim_gene_accession
65          mim_gene_description
66          mirbase_accession
67          mirbase_id
68          protein_id
69          refseq_dna
70          refseq_dna_predicted
71          refseq_peptide
72      refseq_peptide_predicted
73          refseq_genomic
74          rfam
75          unigene
76          uniprot_sptribl
77          uniprot_swissprot_accession
78              wikigene_name
79              wikigene_description
80                  hpa
81                  dbass3_id
82                  dbass3_name
83                  dbass5_id

```

Clone based Ensembl gene
Clone based Ensembl transcript
Clone based VEGA gene
Clone based VEGA transcript

EMBL (Gene)
Ensembl LRG transcript
EntrezGene
VEGA transcript ID(s)
VEGA gene ID(s)
Ensembl transcript (where OTTT shares CDS with ENST)
HAVANA transcript (where ENST shares CDS with OTTT)
HAVANA transcript (where ENST identical to OTTT)

HGNC
HGNC automatic gene
HGNC curated gene
HGNC automatic transcript
HGNC curated transcript
HGNC

MIM
MIM Morbid Affect
MIM Morbid Disease
MIM Gene Affect
MIM Gene Disease
miRBase Accession
miRBase
Protein (Gene)
RefSeq
RefSeq Predicted
RefSeq Predicted Protein
RefSeq Predicted Peptide
RefSeq Genomic

UniProt
UniProt/TrEMBL Accession
UniProt/SwissProt Accession
WikiGene
WikiGene description
Human Protein Atlas Antibody
Database of Aberrant 3' Splice Sites (DBASS3)
DBASS3 Gene
Database of Aberrant 5' Splice Sites (DBASS5)

```

84          dbass5_name           DBASS5 G
85          affy_hc_g110         Affy
86          affy_hg_focus         Affy
87          affy_hg_u133_plus_2    Affy HG U133+
88          affy_hg_u133a_2       Affy HG
89          affy_hg_u133a         Affy HG
90          affy_hg_u133b         Affy HG
91          affy_hg_u95av2        Affy HG
92          affy_hg_u95b          Affy HG
93          affy_hg_u95c          Affy HG
94          affy_hg_u95d          Affy HG
95          affy_hg_u95e          Affy HG
96          affy_hg_u95a          Affy HG
97          affy_hugenefl         Affy HG
98          affy_huex_1_0_st_v2    Affy HuEx 1.0
99          affy_hugene_1_0_st_v1   Affy HuGene 1.0
100         affy_u133_x3p          Affy
101         agilent_cgh_44b        Agilent
102         agilent_wholegenome    Agilent Who
103         codelink               Codelink
104         illumina_humanwg_6_v1   Illumina Human
105         illumina_humanwg_6_v2   Illumina Human
106         illumina_humanwg_6_v3   Illumina Human
107         illumina_humanht_12      Illumina Human
108         phalanx_onearray        Phalanx
109         anatomical_system      Anatomical System (eg
110         development_stage       Development Stage (eg
111         cell_type               Cell Type (eg
112         pathology               Pathology (eg
113         anatomical_system_gnf   Anatomical Syst
114         development_stage_gnf   Development Sta
115         cell_type_gnf           Cell Ty
116         pathology_gnf          Patholo
117         family_description       Ensembl Family Des
118         family                   Ensembl Protein Fami
119         pirsf                   PIRSF SuperF
120         superfamily              Superf
121         smart                   PR
122         profile                 PR
123         prosite                 PR
124         prints                  PR
125         pfam                   PR
126         tigrfam                 TI
127         interpro                Int
128         interpro_short_description Interpro Short Des

```

```

129          interpro_description
130          transmembrane_domain
131          signal_domain
132          ncoils

```

Interpro Des
Transmembran
Signa

We now get a short list of attributes related to the region where the genes are located.

8 Local BioMart databases

The biomaRt package can be used with a local install of a public BioMart database or a locally developed BioMart database and web service. In order for biomaRt to recognize the database as a BioMart, make sure that the local database you create has a name conform with

`database_mart_version`

where database is the name of the database and version is a version number. No more underscores than the ones showed should be present in this name. A possible name is for example

`ensemblLocal_mart_46`

8.1 Minimum requirements for local database installation

More information on installing a local copy of a BioMart database or develop your own BioMart database and webservice can be found on <http://www.biomart.org>. Once the local database is installed you can use biomaRt on this database by:

```
listMarts(host="www.myLocalHost.org", path="/myPathToWebservice/martservice")
mart=useMart("nameOfMyMart",dataset="nameOfMyDataset",host="www.myLocalHost.org", path="/myPathToWebservice/martser
```

For more information on how to install a public BioMart database see:
<http://www.biomart.org/install.html> and follow link databases.

9 Session Info

```
> sessionInfo()
```

```
R version 2.12.0 (2010-10-15)
Platform: x86_64-unknown-linux-gnu (64-bit)

locale:
[1] LC_CTYPE=en_US.UTF-8          LC_NUMERIC=C                  LC_TIME=en_US.UTF-8
[5] LC_MONETARY=C                LC_MESSAGES=en_US.UTF-8      LC_PAPER=en_US.UTF-8
[9] LC_ADDRESS=C                 LC_TELEPHONE=C              LC_MEASUREMENT=en_US.UTF-8

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

other attached packages:
[1] biomaRt_2.6.0

loaded via a namespace (and not attached):
[1] RCurl_1.4-3   XML_3.2-0     tools_2.12.0

> warnings()
NULL
```