

The multiMiR user's guide

Yuanbin Ru, Matt Mulvahill, Spencer Mahaffey, and Katerina Kechris

Package version: multiMiR

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1 Introduction

microRNAs (miRNAs) regulate expression by promoting degradation or repressing translation of target transcripts. miRNA target sites have been cataloged in databases based on experimental validation and computational prediction using a variety of algorithms. Several online resources provide collections of multiple databases but need to be imported into other software, such as R, for processing, tabulation, graphing and computation. Currently available miRNA target site packages in R are limited in the number of databases, types of databases and flexibility.

The R package *multiMiR*, with web server at <http://multimir.ucdenver.edu>, is a comprehensive collection of predicted and validated miRNA-target interactions and their associations with diseases and drugs. *multiMiR* includes several novel features not available in existing R packages:

1. Compilation of 14 different databases, more than any other collection
2. Expansion of databases to those based on disease annotation and drug response, in addition to many experimental and computational databases
3. User-defined cutoffs for predicted binding strength to provide the most confident selection.

The *multiMiR* package enables retrieval of miRNA-target interactions from 14 external databases in R without the need to visit all these databases. Advanced users can also submit SQL queries to the web server to retrieve results. See the [publication on PubMed](#) for additional detail on the database and its creation. The database is now versioned so it is possible to use previous versions of databases from the current R package, however the package defaults to the most recent version.

2 Getting to know the multiMiR database

The *multiMiR* web server <http://multimir.ucdenver.edu> hosts a database containing miRNA-target interactions from external databases. The package *multiMiR* provides functions to communicate with the *multiMiR* web server and its database. The *multiMiR* database is now versioned. By default *multiMiR* will use the most recent version each time *multiMiR* is loaded. However it is now possible to switch between database versions and get information about the *multiMiR* database versions. `multimir_dbInfoVersions()` returns a dataframe with the available versions.

```
library(multiMiR)
## Welcome to multiMiR.
##
## multiMiR database URL has been set to the
## default value: http://multimir.ucdenver.edu/
##
## Database Version: 2.1.0 Updated: 2016-12-22
db.ver = multimir_dbInfoVersions()
db.ver
##   VERSION      UPDATED          RDA      DBNAME      SCHEMA PUBLIC
## 1  2.1.0 2016-12-22 multimir_cutoffs_2.1.rda multimir2_1 multiMiR_DB_schema.sql    1
## 2  2.0.0 2015-05-01    multimir_cutoffs.rda    multimir multiMiR_DB_schema.sql    1
##             TABLES
## 1 multiMiR_dbTables.txt
## 2 multiMiR_dbTables.txt
```

To switch between versions we can use `multimir_switchDBVersion()`.

```
multimir_switchDBVersion(db_version = "2.0.0")
## Now using database version: 2.0.0
multimir_switchDBVersion(db_version = "2.1.0")
## Now using database version: 2.1.0
```

The remaining functions will query the selected version until the package is reloaded or until we switch to another version. Information from each external database is stored in a table in the *multiMiR* database. To see a list of the tables, we can use the `multimir_dbTables()` function.

```
db.tables = multimir_dbTables()
db.tables
## [1] "diana_microt" "elmmo"           "map_counts"     "map_metadata"   "microcosm"      "mir2disease"
## [7] "miranda"       "mirfdb"          "mirecords"     "mirna"         "mirtarbase"     "pharmaco_mir"
## [13] "phenomir"      "pictar"          "pita"          "tarbase"        "target"        "targetscan"
```

To display the database schema, we can use the `multimir_dbSchema()` function. Following is only a portion of the full output.

```
## --
## -- Table structure for table `mirna` --
## --
## 
## DROP TABLE IF EXISTS `mirna`;
## CREATE TABLE `mirna` (
##   mature_mirna_uid INTEGER UNSIGNED AUTO_INCREMENT,      -- mature miRNA unique ID
##   org VARCHAR(4) NOT NULL,                                -- organism abbreviation
##   mature_mirna_acc VARCHAR(20) default NULL,            -- mature miRNA accession
##   mature_mirna_id VARCHAR(20) default NULL,              -- mature miRNA ID/name
##   PRIMARY KEY (mature_mirna_uid),
##   KEY org (org),
```

```

##   KEY mature_mirna_acc (mature_mirna_acc),
##   KEY mature_mirna_id (mature_mirna_id)
## );
##
## --
## -- Table structure for table `target`
## --
## 
## DROP TABLE IF EXISTS `target`;
## CREATE TABLE `target` (
##   target_uid INTEGER UNSIGNED AUTO_INCREMENT,
##   org VARCHAR(4) NOT NULL,
##   target_symbol VARCHAR(80) default NULL,
##   target_entrez VARCHAR(10) default NULL,
##   target_ensembl VARCHAR(20) default NULL,
##   PRIMARY KEY (target_uid),
##   KEY org (org),
##   KEY target_symbol (target_symbol),
##   KEY target_entrez (target_entrez),
##   KEY target_ensembl (target_ensembl)
## );
##
## --
## -- Table structure for table `mirecords`
## --
## 
## DROP TABLE IF EXISTS `mirecords`;
## CREATE TABLE `mirecords` (
##   mature_mirna_uid INTEGER UNSIGNED NOT NULL,
##   target_uid INTEGER UNSIGNED NOT NULL,
##   target_site_number INT(10) default NULL,
##   target_site_position INT(10) default NULL,
##   experiment VARCHAR(160) default NULL,
##   support_type VARCHAR(40) default NULL,
##   pubmed_id VARCHAR(10) default NULL,
##   FOREIGN KEY (mature_mirna_uid)
##     REFERENCES mirna(mature_mirna_uid)
##     ON UPDATE CASCADE ON DELETE RESTRICT,
##   FOREIGN KEY (target_uid)
##     REFERENCES target(target_uid)
##     ON UPDATE CASCADE ON DELETE RESTRICT
## );
##
## .....
## 
## (Please note that only three of the 19 tables are shown here for demonstration
## purpose.)

```

The function `multimir_dbInfo()` will display information about the external miRNA and miRNA-target databases in *multiMiR*, including version, release date, link to download the data, and the corresponding table in *multiMiR*.

```

db.info = multimir_dbInfo()
db.info
##      map_name          source_name source_version  source_date

```

```

## 1 diana_microt           DIANA-microT          5 Sept, 2013
## 2 elmmo                  EIMMo                5 Jan, 2011
## 3 microcosm               MicroCosm            5 Sept, 2009
## 4 mir2disease             miR2Disease          Mar 14, 2011
## 5 miranda                 miRanda              Aug, 2010
## 6 mirdb                  miRDB                5 Aug, 2014
## 7 mirecords               miRecords             4 Apr 27, 2013
## 8 mirtarbase              miRTarBase            6.1 Sept, 2015
## 9 pharmaco_mir             Pharmaco-miR (Verified Sets)
## 10 phenomir                PhenomiR             2 Feb 15, 2011
## 11 pictar                  PicTar               2 Dec 21, 2012
## 12 pita                    PITA                 6 Aug 31, 2008
## 13 tarbase                 TarBase              6 2012
## 14 targetscan              TargetScan            6.2 Jun, 2012
##
# source_url
## 1 http://diana.imis.athena-innovation.gr/DianaTools/index.php?r=miroCDS/index
## 2 http://www.mirz.unibas.ch/miRNATargetPredictionBulk.php
## 3 http://www.ebi.ac.uk/enright-srv/microcosm/cgi-bin/targets/v5/download.pl
## 4 http://www.mir2disease.org
## 5 http://www.microrna.org/microrna/getDownloads.do
## 6 http://mirdb.org
## 7 http://mirecords.biolead.org/download.php
## 8 http://mirtarbase.mbc.nctu.edu.tw/php/download.php
## 9 http://www.pharmaco-mir.org/home/download_VERSE_db
## 10 http://mips.helmholtz-muenchen.de/phenomir/
## 11 http://dorina.mdc-berlin.de
## 12 http://genie.weizmann.ac.il/pubs/mir07/mir07_data.html
## 13 http://diana.imis.athena-innovation.gr/DianaTools/index.php?r=tarbase/index
## 14 http://www.targetscan.org/cgi-bin/targetscan/data_download.cgi?db=vert_61

```

Among the 14 external databases, eight contain predicted miRNA-target interactions (DIANA-microT-CDS, EIMMo, MicroCosm, miRanda, miRDB, PicTar, PITA, and TargetScan), three have experimentally validated miRNA-target interactions (miRecords, miRTarBase, and TarBase) and the remaining three contain miRNA-drug/disease associations (miR2Disease, Pharmaco-miR, and PhenomiR). To check these categories and databases from within R, we have a set of four helper functions:

```

predicted_tables()
## [1] "diana_microt" "elmmo"      "microcosm"   "miranda"    "mirdb"      "pictar"
## [7] "pita"         "targetscan"

validated_tables()
## [1] "mirecords"  "mirtarbase"  "tarbase"

diseasedrug_tables()
## [1] "mir2disease" "pharmaco_mir" "phenomir"

reverse_table_lookup("targetscan")
## [1] "predicted"

```

To see how many records are in these 14 external databases we refer to the `multimir_dbCount` function.

```

db.count = multimir_dbCount()
db.count
##      map_name human_count mouse_count rat_count total_count
## 1 diana_microt     7664602     3747171        0    11411773
## 2 elmmo            3959112     1449133     547191    5955436
## 3 microcosm         762987     534735     353378    1651100
## 4 mir2disease       2875          0          0      2875

```

```

## 5      miranda    5429955    2379881    247368    8057204
## 6      mirdb      1124831    654192     195136    1974159
## 7      mirecords   2425       449        171       3045
## 8      mirtarbase  440040    48731      470       489281
## 9      pharmaco_mir 308        5          0         313
## 10     phenomir    15138      491        0         15629
## 11     pictar      404066    302236     0         706302
## 12     pita        7710936    5163153     0         12874089
## 13     tarbase     23447      9024       67        32538
## 14     targetscan   6298122    2590814     0         8888936
apply(db.count[,-1], 2, sum)
## human_count mouse_count rat_count total_count
##      33838844     16880015     1343781     52062680

```

The current version of *multiMiR* contains nearly 50 million records.

3 List miRNAs, genes, drugs and diseases in the multiMiR database

In addition to functions displaying database and table information, the *multiMiR* package also provides the `list.multimir()` function to list all the unique miRNAs, target genes, drugs, and diseases in the *multiMiR* database. An option for limiting the number of returned records has been added to help with testing and exploration.

```

miRNAs  = list.multimir("mirna", limit = 10)
genes   = list.multimir("gene", limit = 10)
drugs   = list.multimir("drug", limit = 10)
diseases = list.multimir("disease", limit = 10)
# executes 2 separate queries, giving 20 results
head(miRNAs)
##   mature_mirna_uid org mature_mirna_acc mature_mirna_id
## 1                 936 hsa
## 2                 1284 hsa           hsa-let-7
## 3                 202 hsa           hsa-let-7f1
## 4                 815 hsa           hsa-let-7a-1
## 5                 816 hsa           hsa-let-7a-2
## 6                 817 hsa           hsa-let-7a-3
head(genes)
##   target_uid org target_symbol target_entrez target_ensembl
## 1       60954 hsa
## 2       60956 hsa           ENSG00000011177
## 3       60957 hsa           ENSG00000051415
## 4       60958 hsa           ENSG00000083622
## 5       60959 hsa           ENSG00000099964
## 6       60960 hsa           ENSG00000100101
head(drugs)
##             drug
## 1 3,3'-diindolylmethane
## 2      5-fluorouracil
## 3          abt-737
## 4      alitretinoin
## 5      arabinocytosine
## 6      arsenic trioxide
head(diseases)
##             disease

```

1 ACTH-INDEPENDENT MACRONODULAR ADRENAL HYPERPLASIA; AIMAH
2 ACUTE LYMPHOBLASTIC LEUKEMIA (ALL)
3 ACUTE MYELOGENOUS LEUKEMIA (AML)
4 ACUTE MYELOID LEUKEMIA (AML)
5 ACUTE PROMYELOCYTIC LEUKEMIA (APL)
6 ADENOMA

The current version of *multiMiR* has 5830 miRNAs and 97186 target genes from human, mouse, and rat, as well as 64 drugs and 223 disease terms. Depending on the speed of your Internet connection, it may take a few minutes to retrieve the large number of target genes.

4 Use `get.multimir()` to query the multiMiR database

`get.multimir()` is the main function in the package to retrieve predicted and validated miRNA-target interactions and their disease and drug associations from the *multiMiR* database.

To get familiar with the parameters in `get.multimir()`, you can type `?get.multimir` or `help(get.multimir)` in R. In the next section, many examples illustrate the use of the parameters.

5 Examples of multiMiR queries

In this section a variety of examples are described on how to query the multiMiR database.

5.1 Example 1: Retrieve all validated target genes of a given miRNA

In the first example, we ask what genes are validated targets of hsa-miR-18a-3p.

```

# The default is to search validated interactions in human
example1 <- get.multimir(mirna = 'hsa-miR-18a-3p', summary = TRUE)
## Searching mirecords ...
## Searching mirtarbase ...
## Searching tarbase ...
names(example1)
## [1] "validated"      "predicted"      "disease.drug"   "queries"       "summary"
# Detailed information of the validated miRNA-target interaction
head(example1$validated)
##           database mature_mirna_acc mature_mirna_id target_symbol target_entrez target_ensembl
## 1      mirecords      MIMAT0002891  hsa-miR-18a-3p          KRAS            3845 ENSG00000133703
## 2     mirtarbase      MIMAT0002891  hsa-miR-18a-3p          ACLY             47 ENSG00000131473
## 3     mirtarbase      MIMAT0002891  hsa-miR-18a-3p          AP1B1            162 ENSG00000100280
## 4     mirtarbase      MIMAT0002891  hsa-miR-18a-3p          ARF5            381 ENSG00000004059
## 5     mirtarbase      MIMAT0002891  hsa-miR-18a-3p          ATP2B4            493 ENSG00000058668
## 6     mirtarbase      MIMAT0002891  hsa-miR-18a-3p          ATP5A1            498 ENSG00000152234
##           experiment support_type pubmed_id
## 1 Western blot//Luciferase activity assay           19372139
## 2                               CLASH Functional MTI (Weak)  23622248
## 3                               CLASH Functional MTI (Weak)  23622248
## 4                               CLASH Functional MTI (Weak)  23622248
## 5                               HITS-CLIP Functional MTI (Weak) 19536157
## 6                               CLASH Functional MTI (Weak)  23622248

```

```
# Which interactions are supported by Luciferase assay?
example1$validated[grep("Luciferase", example1$validated[, "experiment"]), ]
##      database mature_mirna_acc mature_mirna_id target_symbol target_entrez target_ensembl
## 1      mirecords      MIMAT0002891   hsa-miR-18a-3p          KRAS           3845 ENSG00000133703
## 52     mirtarbase      MIMAT0002891   hsa-miR-18a-3p          KRAS           3845 ENSG00000133703
##                                         experiment support_type pubmed_id
## 1             Western blot//Luciferase activity assay           19372139
## 52 Luciferase reporter assay//qRT-PCR//Western blot Functional MTI 19372139
example1$summary[example1$summary[, "target_symbol"] == "KRAS", ]
##      mature_mirna_acc mature_mirna_id target_symbol target_entrez target_ensembl mirecords mirtarbase
## 1      MIMAT0002891   hsa-miR-18a-3p          KRAS           3845 ENSG00000133703           1           1
##      validated.sum all.sum
## 1              2       2
```

It turns out that *KRAS* is the only target validated by Luciferase assay. The interaction was recorded in miRecords and miRTarBase and supported by the same literature, whose PubMed ID is in column `pubmed_id`. The summary (by setting `summary = TRUE` when calling `get.multimir()`) shows the number of records in each of the external databases and the total number of databases supporting the interaction.

5.2 Example 2: Retrieve miRNA-target interactions associated with a given drug or disease

In this example we would like to know which miRNAs and their target genes are associated with Cisplatin, a chemotherapy drug used in several cancers.

```
example2 <- get.multimir(disease.drug = 'cisplatin', table = 'disease.drug')
## Searching mir2disease ...
## Searching pharmaco_mir ...
## Searching phenomir ...
names(example2)
## [1] "validated"      "predicted"      "disease.drug"    "queries"        "summary"
nrow(example2$disease.drug)
## [1] 45
head(example2$disease.drug)
##      database mature_mirna_acc mature_mirna_id target_symbol target_entrez target_ensembl
## 1 pharmaco_mir      MIMAT0000772   hsa-miR-345-5p          ABCC1           4363 ENSG00000103222
## 2 pharmaco_mir      MIMAT0000720   hsa-miR-376c-3p          ALK7
## 3 pharmaco_mir      MIMAT0000423   hsa-miR-125b-5p          BAK1           578 ENSG00000030110
## 4 pharmaco_mir                               hsa-miR-34          BCL2           596 ENSG00000171791
## 5 pharmaco_mir      MIMAT0000318   hsa-miR-200b-3p          BCL2           596 ENSG00000171791
## 6 pharmaco_mir      MIMAT0000617   hsa-miR-200c-3p          BCL2           596 ENSG00000171791
##      disease_drug paper_pubmedID
## 1      cisplatin      20099276
## 2      cisplatin      21224400
## 3      cisplatin      21823019
## 4      cisplatin      18803879
## 5      cisplatin      21993663
## 6      cisplatin      21993663
```

`get.multimir()` returns 53 miRNA-target pairs. For more information, we can always refer to the published papers with PubMed IDs in column `paper_pubmedID`.

5.3 Example 3: Select miRNAs predicted to target a gene

`get.multimir()` also takes target gene(s) as input. In this example we retrieve miRNAs predicted to target *Gnb1* in mouse. For predicted interactions, the default is to query the top 20% predictions within each external database, which is equivalent to setting parameters `predicted.cutoff = 20` and `predicted.cutoff.type = 'p'` (for percentage cutoff). Here we search the top 35% among all conserved and nonconserved target sites.

```
example3 <- get.multimir(org      = "mmu",
                         target    = "Gnb1",
                         table     = "predicted",
                         summary   = TRUE,
                         predicted.cutoff      = 35,
                         predicted.cutoff.type = "p",
                         predicted.site        = "all")

## Searching diana_microt ...
## Searching elmmo ...
## Searching microcosm ...
## Searching miranda ...
## Searching mirdb ...
## Searching pictar ...
## Searching pita ...
## Searching targetscan ...
names(example3)
## [1] "validated"      "predicted"      "disease.drug"   "queries"       "summary"
head(example3$predicted)
##           database mature_mirna_acc  mature_mirna_id target_symbol target_entrez target_ensembl
## 1 diana_microt      MIMAT0000663 mmu-miR-218-5p      Gnb1          14688 ENSMUSG00000029064
## 2 diana_microt      MIMAT0017276 mmu-miR-493-5p      Gnb1          14688 ENSMUSG00000029064
## 3 diana_microt      MIMAT0000656 mmu-miR-139-5p      Gnb1          14688 ENSMUSG00000029064
## 4 diana_microt      MIMAT0014946 mmu-miR-3074-2-3p     Gnb1          14688 ENSMUSG00000029064
## 5 diana_microt      MIMAT0000144 mmu-miR-132-3p      Gnb1          14688 ENSMUSG00000029064
## 6 diana_microt      MIMAT0020608 mmu-miR-5101       Gnb1          14688 ENSMUSG00000029064
##   score
## 1 0.975
## 2 0.964
## 3 0.96
## 4 0.921
## 5 0.92
## 6 0.918
head(example3$summary)
##   mature_mirna_acc mature_mirna_id target_symbol target_entrez target_ensembl diana_microt
## 1      MIMAT0000133 mmu-miR-101a-3p      Gnb1          14688 ENSMUSG00000029064      1
## 2      MIMAT0000616 mmu-miR-101b-3p      Gnb1          14688 ENSMUSG00000029064      1
## 3      MIMAT0003476 mmu-miR-669b-5p      Gnb1          14688 ENSMUSG00000029064      1
## 4      MIMAT0000663 mmu-miR-218-5p      Gnb1          14688 ENSMUSG00000029064      1
## 5      MIMAT0002106 mmu-miR-465a-5p      Gnb1          14688 ENSMUSG00000029064      1
## 6      MIMAT0003739 mmu-miR-673-5p      Gnb1          14688 ENSMUSG00000029064      1
##   elmmo microcosm miranda mirdb pictar pita targetscan predicted.sum all.sum
## 1     1       1     1     0     2     0       1       6       6
## 2     1       1     1     0     2     0       1       6       6
## 3     0       0     1     2     0     1       1       5       5
## 4     2       0     0     0     0     2       2       4       4
## 5     1       0     1     0     0     0       2       4       4
## 6     0       0     0     0     2     1       1       4       4
```

The records in `example3$predicted` are ordered by scores from best to worst within each external database. Once again, the summary option allows us to examine the number of target sites predicted by each external database and the total number of databases predicting the interaction.

Finally we examine how many predictions each of the databases has.

```
apply(example3$summary[, 6:13], 2, function(x) sum(x > 0))
## diana_microt      elmmo     microcosm    miranda      mirdb      pictar      pita
##      105          51          5          44          16          9         132
## targetscan
##      58
```

5.4 Example 4: Select miRNA(s) predicted to target most, if not all, of the genes of interest

You may have a list of genes involved in a common biological process. It is interesting to check whether some, or all, of these genes are targeted by the same miRNA(s). Here we have four genes involved in chronic obstructive pulmonary disease (COPD) in human and want to know what miRNAs target these genes by searching the top 500,000 predictions in each external database.

```
example4 <- get.multimir(org      = 'hsa',
                         target    = c('AKT2', 'CERS6', 'S1PR3', 'SULF2'),
                         table     = 'predicted',
                         summary   = TRUE,
                         predicted.cutoff.type = 'n',
                         predicted.cutoff      = 500000)
## Number predicted cutoff (predicted.cutoff) 500000 is larger than the total number of records in table p
## Searching diana_microt ...
## Searching elmmo ...
## Searching microcosm ...
## Searching miranda ...
## Searching mirdb ...
## Searching pictar ...
## Searching pita ...
## Searching targetscan ...
```

Then we count the number of target genes for each miRNA.

```
example4.counts <- addmargins(table(example4$summary[, 2:3]))
example4.counts <- example4.counts[-nrow(example4.counts), ]
example4.counts <- example4.counts[order(example4.counts[, 5], decreasing = TRUE), ]
head(example4.counts)
##                  target_symbol
## mature_miRNA_id AKT2 CERS6 S1PR3 SULF2 Sum
## hsa-miR-429     1     1     0     1     3
## hsa-miR-524-5p  1     1     0     1     3
## hsa-miR-548p    0     1     1     1     3
## hsa-miR-876-5p  0     1     1     1     3
## hsa-let-7a-5p   1     0     0     1     2
## hsa-let-7b-5p   1     0     0     1     2
```

5.5 Example 5: Retrieve interactions between a set of miRNAs and a set of genes

In this example, we profiled miRNA and mRNA expression in poorly metastatic bladder cancer cell lines T24 and Luc, and their metastatic derivatives FL4 and LuL2, respectively. We identified differentially expressed miRNAs and genes between the metastatic and poorly metastatic cells. Let's load the data.

```
load(url("http://multimir.ucdenver.edu/bladder.rda"))
```

Variable DE.miRNA.up contains 9 up-regulated miRNAs and variable DE.entrez.dn has 47 down-regulated genes in the two metastatic cell lines. The hypothesis is that interactions between these miRNAs and genes whose expression changed at opposite directions may play a role in cancer metastasis. So we use multiMiR to check whether any of the nine miRNAs could target any of the 47 genes.

```
## Searching diana_microt ...
## Searching elmmo ...
## Searching microcosm ...
## Searching miranda ...
## Searching mirdb ...
## Searching pictar ...
## Searching pita ...
## Searching targetscan ...
## Searching mirecords ...
## Searching mirtarbase ...
## Searching tarbase ...
## Searching mir2disease ...
## Searching pharmaco_mir ...
## Searching phenomir ...
```

In the result, there are 3 unique miRNA-gene pairs that have been validated.

```
##      database mature_mirna_acc mature_mirna_id target_symbol target_entrez
## 1 mirtarbase      MIMAT0000087   hsa-miR-30a-5p        FDX1        2230
## 2 mirtarbase      MIMAT0000087   hsa-miR-30a-5p        LIMCH1      22998
## 3 tarbase        MIMAT0000087   hsa-miR-30a-5p        FDX1        2230
## 4 tarbase        MIMAT0000424   hsa-miR-128         NEK2        4751
## 5 tarbase        MIMAT0000087   hsa-miR-30a-5p        LIMCH1      22998
##      target_ensembl          experiment support_type pubmed_id
## 1 ENSG00000137714          Proteomics Functional MTI (Weak) 18668040
## 2 ENSG00000064042 pSILAC//Proteomics;Other Functional MTI (Weak) 18668040
## 3 ENSG00000137714          Proteomics           positive
## 4 ENSG00000117650          Microarray            positive
## 5 ENSG00000064042          Proteomics           positive
```

Two miRNAs are associated with bladder cancer in miR2Disease and PhenomiR.

```
##      database mature_mirna_acc mature_mirna_id target_symbol target_entrez
## 18  mir2disease      MIMAT0000418   hsa-miR-23b-3p        NA        NA
## 711  phenomir        MIMAT0000418   hsa-miR-23b-3p        NA        NA
## 311  phenomir        MIMAT0000449   hsa-miR-146a-5p        NA        NA
##      target_ensembl disease_drug
## 18                 NA bladder cancer
## 711                 NA Bladder cancer
## 311                 NA Bladder cancer
##                                     paper_pubmedID
## 18  2007. Micro-RNA profiling in kidney and bladder cancers.
## 711
## 311
```

The predicted databases predict 65 miRNA-gene pairs between the 9 miRNAs and 28 of the 47 genes.

```
## [1] 9
## [1] 28
unique.pairs <-
  unique(data.frame(miRNA.ID = as.character(example5$predicted$mature_mirna_id),
                    target.Entrez = as.character(example5$predicted$target_entrez)))
nrow(unique.pairs)
## [1] 65
##           miRNA.ID target.Entrez
## 1 hsa-miR-182-5p      2017
## 2 hsa-miR-182-5p      1112
## 3 hsa-miR-30d-5p     22998
## 4 hsa-miR-30a-5p     22998
## 5 hsa-miR-30b-5p     22998
## 6 hsa-miR-182-5p      5962
```

Results from each of the predicted databases are already ordered by their scores from best to worst.

6 Direct query to the database on the multiMiR web server

As shown previously, `get.multimir` is the main function to retrieve information from the *multiMiR* database, which is hosted at <http://multimir.ucdenver.edu>. The function builds one SQL query for every external database that the user is going to search, submits the query to the web server, and parses, combines, and summarizes results from the web server. For advanced users, there are a couple ways to query the *multiMiR* database without using the *multiMiR* package; but they have to be familiar with SQL queries. In general, users are still advised to use the `get.multimir()` function when querying multiple external databases in *multiMiR*.

6.1 Direct query on the web server

The *multiMiR* package communicates with the *multiMiR* database via the script http://multimir.ucdenver.edu/cgi-bin/multimir_univ.pl on the web server. Once again, data from each of the external databases is stored in a table in *multiMiR*. There are also tables for miRNAs (table *mirna*) and target genes (table *target*).

NOTE: While it is possible to complete short queries from a browser, the limits of submitting a query through typing in the address bar of a browser are quickly reached (8192 characters total). If you are a developer you should use your preferred method to submit a HTTP POST which will allow for longer queries. The fields to include are `query` and `dbName`. `query` is the SQL query to submit. `dbName` is the DBNAME column from a call to `multimir_dbInfoVersions()`, however if this is excluded the current version is the default.

To learn about the structure of a table (e.g. DIANA-microT data in table *diana_microt*), users can use URL

http://multimir.ucdenver.edu/cgi-bin/multimir_univ.pl?query=describe+diana_microt

Similar with Example 1, the following URL searches for validated target genes of hsa-miR-18a-3p in miRecords.

[http://multimir.ucdenver.edu/cgi-bin/multimir.pl?query=SELECT+m.mature_mirna_acc,+m.mature_mirna_id,t.target_symbol,+t.target_entrez,+t.target_ensembl,i.experiment,i.support_type,i.pubmed_id+FROM+mirna+AS+m+INNER+JOIN+mirecords+AS+i+INNER+JOIN+target+AS+t+ON+\(m.mature_mirna_uid=i.mature_mirna_uid+and+i.target_uid=t.target_uid\)+WHERE+m.mature_mirna_id='hsa-miR-18a-3p'](http://multimir.ucdenver.edu/cgi-bin/multimir.pl?query=SELECT+m.mature_mirna_acc,+m.mature_mirna_id,t.target_symbol,+t.target_entrez,+t.target_ensembl,i.experiment,i.support_type,i.pubmed_id+FROM+mirna+AS+m+INNER+JOIN+mirecords+AS+i+INNER+JOIN+target+AS+t+ON+(m.mature_mirna_uid=i.mature_mirna_uid+and+i.target_uid=t.target_uid)+WHERE+m.mature_mirna_id='hsa-miR-18a-3p')

As you can see, the query is long and searches just one of the three validated tables in *multiMiR*. While in Example 1, one line of R command using the `get.multimir()` function searches, combines and summarizes results from all three validated external databases (miRecords, miRTarBase and TarBase).

6.2 Direct query in R

The same direct queries we did above on the web server can be done in R as well. This is the preferred method if you are unfamiliar with HTTP POST. Be sure to set the correct database version, if you wish to change versions, before calling `search.multimir()` it uses the currently set version.

To show the structure of table *diana_microt*:

```
direct2 <- search.multimir(query = "describe diana_microt")
direct2
##           Field      Type Null Key Default Extra
## 1 mature_mirna_uid int(10) unsigned    NO MUL
## 2     target_uid int(10) unsigned    NO MUL
## 3       miTG_score    double    NO MUL
## 4        UTR3_hit int(10) unsigned    NO
## 5        CDS_hit int(10) unsigned    NO
```

To search for validated target genes of hsa-miR-18a-3p in miRecords:

```
qry <- "SELECT m.mature_mirna_acc, m.mature_mirna_id, t.target_symbol,
         t.target_entrez, t.target_ensembl, i.experiment, i.support_type,
         i.pubmed_id
      FROM mirna AS m INNER JOIN mirecords AS i INNER JOIN target AS t
      ON (m.mature_mirna_uid=i.mature_mirna_uid and
          i.target_uid=t.target_uid)
      WHERE m.mature_mirna_id='hsa-miR-18a-3p'"
direct3 <- search.multimir(query = qry)
direct3
##   mature_mirna_acc mature_mirna_id target_symbol target_entrez target_ensembl
## 1      MIMAT0002891      hsa-miR-18a-3p          KRAS            3845 ENSG00000133703
##                                         experiment support_type pubmed_id
## 1 Western blot//Luciferase activity assay                  19372139
```

7 Session Info

```
sessionInfo()
## R version 3.3.2 (2016-10-31)
## Platform: x86_64-apple-darwin13.4.0 (64-bit)
## Running under: macOS Sierra 10.12.4
##
## locale:
## [1] en_US.UTF-8/en_US.UTF-8/en_US.UTF-8/C/en_US.UTF-8/en_US.UTF-8
##
## attached base packages:
## [1] stats      graphics   grDevices  utils      datasets   methods    base
##
## other attached packages:
## [1] multiMiR_0.99.3 knitr_1.15.1    BiocStyle_2.2.1
```

```
##  
## loaded via a namespace (and not attached):  
## [1] Rcpp_0.12.10    XML_3.98-1.6   digest_0.6.12   rprojroot_1.2   bitops_1.0-6    backports_1.0.5  
## [7] magrittr_1.5    evaluate_0.10  stringi_1.1.5  rmarkdown_1.4   tools_3.3.2     stringr_1.2.0  
## [13] purrrr_0.2.2   RCurl_1.95-4.8  yaml_2.1.14    htmltools_0.3.5  
warnings()  
## NULL
```