rGriffin

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Introduction

Phenotype is the result of the interaction of multiple genes

Boolean syncrolous networks are a simplification, proposed by kaufman nodes interactions state attractors Biological problem

We can approximate the interaction graph and the attractors We are not sure of the functions

Does the solution exist? Is there only one solution?

Using a simple file interface, griffin takes as inputs biologically meaningful constraints and turns them into a symbolic representation. Using a SAT engine, griffin explores the Boolean Network search space, finding all satisfying assignments that are compatible with the specified constraints

This allows to determine if the information is sufficient Find the models that satisfy the restrictions

Incorporate multiple types of biological data with different levels of certanty.

Search for patterns in the families of possible networks

IGriffin can be found in: http://turing.iimas.unam.mx/griffin/

Queries

We are not iterating all possible networks, but determining the networks that satisfy our conditions.

If there are more restrictions we expect less networks. We recommend beggining with more stringent queries and then relaxing the restrictions.

The time that griffin will take depends on:

- Number of possible solutions
- The number of regulators of each node

Create a query

All queries start with a topology that describes the nodes and its interactions. The function create.gquery.graph() takes a dataframe with columns for: source node, target node, and type of interaction.

It also takes a vector with the node names.

Depending on the sign the interactions can **positive** or **negative**. If in every condition the regulation will have the same sign the interaction is **ambiguous**. However, if you are not sure if the regulation is positive or negative in all contexts you can say that the interaction is **ambiguous**. Depending on the degree of confidence in the existence of the interaction, the interactions can be: **mandatory** if you are sure the interaction will happen or **optional** if you suspect the interaction exists but you are not sure.

The valid types of interctions are:

- false: Contradiction
- MA: Mandatory, ambiguous
- MPU (or +): Mandatory, positive, unambiguous
- MPPA: Mandatory, positive, possibly ambiguous
- MNU (or -): Mandatory, negative, unambiguous
- MNPA: Mandatory, negative, possibly ambiguous
- MUSU: Mandatory, unknown sign, unambiguous
- MUSPA: Mandatory, unknown sign, possibly ambiguous
- · NR: No regulation
- OA: Optional, ambiguous
- o OPU: Optional, positive, unambiguous
- OPPA: Optional, positive, possibly ambiguous
- · ONU: Optional, negative, unambiguous
- · ONPA: Optional, negative, possibly ambiguous
- o OUSU: Optional, unknown sign, unambiguous
- true: Tautology

For example, suppose a network where:

- a activates b
- b and c inhibit each other
- **b** and **c may** have a positive self-regulatory loops.

We can codify this information as:

```
genes = c('a', 'b', 'c')
inter = data.frame(source=c('a', 'b', 'b', 'c', 'c'),
                  target=c('b','b','c','b','c'),
                   type=c('+','OPU','-','-','OPU'),
                    stringsAsFactors = F )
inter
#>
     source target type
#> 2
          b
                 b OPU
#> 3
          b
#> 4
          С
                 b
                 c OPU
#> 5
          C
```

We then create the query q. This creates an instance of the class query in the JVM.

```
q = create.gquery.graph(inter, genes)
q
#> [1] "Java-Object{mx.unam.iimas.griffin.r.RGriffinQuery@1bba400}"
```

If you want to see the query use the print method.

```
print(q)
#> Gene Regulation Specification
#> Set of genes: [a, b, c]
#> geneOrder: null
#> Known regulations: None
#> Hypothetical regulations: None
#> Gene Interactions: [GeneInteraction [source=a, target=b, type=MPU], GeneInteraction
[source=b, target=b, type=OPU], GeneInteraction [source=b, target=c, type=MNU],
GeneInteraction [source=c, target=b, type=MNU], GeneInteraction [source=c, target=c,
type=OPU]]
#> Stable states: None
#> Prohibited states: null
#> Known Regulators: None
#> Hypothetical Regulators: None
#> Extended regulators: None
#> Known Regulations: None
#> Hypothetical Regulations: None
#> Cyclic attractors: None
#> Input-output pairs: []
#> Subspaces: None
#> Mutant experiments: None
#> State transition rules: None
#> exactRegulations: false
#> additionalSteadyStates: false
#> divideQueryByTopology: false
#> topologyIteratorType: SEQUENTIAL
#> limitTopologyRange: false
#> limitToNumberOfNetworks: -1
#> topologyRange: None
#> topologicalDistanceRadius: 1
#> blockSteadyAPosteriory: false
#> additonalCycles: false
#> ambiguityNotAllowed: true
#> numberOfCycles:0
```

Add restrictions to the query

It is possible to add more restrictions to the query. It is important to remember that if there are more restrictions we expect less networks.

For example, suppose that we have some information about the expected cell types. We can add this restrictions as attractors.

For example, suppose that we know that the attractors are:

- Only express b => c(0,1,0)
- Only express c => c(0,0,1)

We can also add partial attractors where we lack information

• Do not express **a** or **c** but we have NO information or $b \Rightarrow c(0, x, 0)$

We codify this information as:

We can add this information to the query with the add.gquery.attractors() function:

```
q = add.gquery.attractors(q, attr)
```

You can add aditional restrictions like:

- add.gquery.cycle() Add target cycle
- add.gquery.mutant() Add mutant with attractors
- add.gquery.prohibited.attractors() Add prohibited attractors
- add.gquery.transition() Add a transition between two succesive states
- add.gquery.trapspace() Add a trapspace

Run the query

Once you have created the query with run.gquery(). This function will return all the networks that satisfy the restrictions

```
nets = run.gquery(q)
print(nets)
#> [1] "targets, factors\na, false\nb, ((((!a&b)&!c)|((a&!b)&!c))|((a&b)&!c))\nc, (!b&c)\n"
#> [2] "targets, factors\na, false\nb, (((((!a&b)&!c)|((a&!b)&!c))|((a&b)&!c))|((a&b)&c))\nc, (!b&c)\n"
#> [3] "targets, factors\na, false\nb, ((((((!a&b)&!c)|((!a&b)&c))|((a&!b)&!c))|((a&b)&!c))|((a&b)&c))\nc, (!b&c)\n"
#> [4] "targets, factors\na, false\nb, ((((!a&b)&!c)|((a&b)&!c))|((a&b)&c))\nc, (!b&c)\n"
#> [5] "targets, factors\na, false\nb, (((((!a&b)&!c)|((a&!b)&!c))|((a&!b)&c))|((a&b)&!c))|((a&b)&c))\nc, (!b&c)\n"
```

The function run.gquery() includes multiple options that can be seen in the documentation. Some of the most important are:

- allow.hypothesis activate or deactivate hypothetical regulations
- allow.additional.states allows networks with additional fixed-point attractors to those specified in the query
- allow.additional.cycles allows networks with additional cyclic attractors to those specified in the query
- return.network.limit limit the maximum number of networks that the query will return

Connect to other packages

BoolNet

The R package includes various functions to import and export data to BoolNet.

It is possible to obtain the topology of a BooleanNetwork object using get.net.topology(). This function determines the sign of each regulation as positive '+', negative '-' or ambiguous 'MA'. All regulations between nodes are considered mandatory. The function can also detect Non-functional regulations 'NR'.

```
library(BoolNet)
data("cellcycle")
topology <- get.net.topology(cellcycle)</pre>
topology
#>
      Source Target Interaction
#> 1
        CycD
               CycD
#> 2
        CycD
                 Rb
#> 3
        CycE
                 Rb
#> 4
        CycA
                 Rb
         p27
                 Rb
#> 5
        CycB
                 Rb
#> 6
                E2F
#> 7
          Rb
#> 8
        CycA
                E2F
#> 9
         p27
                E2F
                E2F
#> 10
        CycB
#> 11
          Rb
               CycE
#> 12
         E2F
               CycE
#> 13
          Rb
               CycA
#> 14
         E2F
               CycA
#> 15
        CycA
               CycA
#> 16
       Cdc20
               CycA
#> 17
        Cdh1
               CycA
#> 18 UbcH10
               CycA
#> 19
        CycD
                p27
#> 20
                p27
        CycE
#> 21
        CycA
                p27
#> 22
                p27
         p27
#> 23
        CycB
                p27
#> 24
        CycB
              Cdc20
#> 25
        CycA
               Cdh1
#> 26
         p27
               Cdh1
#> 27
       Cdc20
               Cdh1
#> 28
        CycB
               Cdh1
#> 29
        CycA UbcH10
#> 30 Cdc20 UbcH10
#> 31
        Cdh1 UbcH10
#> 32 UbcH10 UbcH10
#> 33
        CycB UbcH10
#> 34
       Cdc20
               CycB
#> 35
        Cdh1
               CycB
```

It is also possible to convert an AttractorInfo object into a data.frame using attractor2dataframe.

```
cc.attr <- getAttractors(cellcycle)
cycle <- attractor2dataframe(cc.attr)</pre>
```

```
cycle <- cycle[cellcycle$genes] #remove info columns
cycle
    CycD Rb E2F CycE CycA p27 Cdc20 Cdh1 UbcH10 CycB
#> 1
       0 1
                  0
                      0
                          1
#> 2
       1 0
             0
                  1
                       1
                          0
                                0
                                     0
                                           0
                                                0
#> 3
       1 0
             0
                  0
                      1
                          0
                                           1
                                                1
#> 4
       1 0
             0
                  0
                      1
                          0
                                1
                                     0
                                           1
                                                1
#> 5
       1 0 0
                  0
                      0
                          0
                                1
                                    1
                                           1
       1 0 1
                  0
#> 7
       1 0
             1
                  1
                      0
                          0
                                0
                                    1
                                           0
                                                0
#> 8
       1 0
             1
                  1
                       1
                          0
                                0
                                     1
                                           0
                                                0
```

Using this functions, it is possible to create a query that corresponds to a Boolean network. As we expect a cyclic attractor we will use add.gquery.cycle().

```
q <- create.gquery.graph(topology, cellcycle$genes)
q <- add.gquery.cycle(q, cycle)</pre>
```

It is also possible to export the networks generated by run.gquery() directly to BoolNet with the option return = "BoolNet". This option generates an iterator object, that returns the BooleanNetwork objects one by one using the function iterators::nextElem().

For example, if we use the three node example network:

```
nets = run.gquery(q,return = "BoolNet")
#> Warning in boolnet.model.iterator(controller): iterators is not attached
iterators::nextElem(nets)
#> Boolean network with 3 genes
#>
#> Involved genes:
#> a b c
#>
#> Transition functions:
#> a = 0
#> b = ((((!a&b)&!c))((a&!b)&!c)))((a&b)&!c))
#> c = (!b&c)
#>
#> Knocked-out and over-expressed genes:
#> a = 0
```

If there are no more available networks the <code>nextElem()</code> method will rise an error: <code>Error</code> in <code>obj\$nextElem()</code> : <code>StopIteration</code>.

Other

It is possible to plot the network topology dataframe with the R package <code>igraph</code>. This dataframe can also be used to import and export the network topology to other resources like the python library <code>networkx</code> or to the software <code>Cytoscape</code>.

It is possible to export the network functions as an SBML file using the BoolNet function toSBML().

References

Muñoz, S., Carrillo, M., Azpeitia, E., & Rosenblueth, D. A. (2018). Griffin: A Tool for Symbolic Inference of Synchronous Boolean Molecular Networks. Frontiers in genetics, 9, 39.