# Model manipulation

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#### INTRODUCTION

In this tutorial, we will do a manipulation with the simple model of the first few reactions of the glycolysis metabolic pathway as created in the "Model Creation" tutorial.

Glycolysis is the metabolic pathway that occurs in most organisms in the cytosol of the cell. First, we will use the beginning of that pathway to create a simple constraint-based metabolic network (Figure 1).

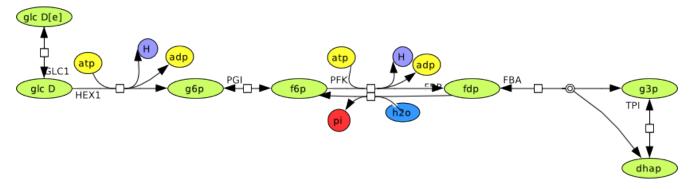


Figure 1: A small metabolic network consisting of the seven reactions in the glycolysis pathway.

At the beginning of the reconstruction, the initial step is to assess an integrity of the draft reconstruction. Furthermore, an accuracy evaluation is needed: check necessity of each reaction and metabolite, the accuracy of the stoichiometry and direction and reversibility of the reactions.

After creating or loading the model, the model can be modified to simulate different conditions, such as:

- Creating, adding, and handling reactions;
- · Adding Exchange, Sink and Demand reactions;
- · Altering reaction bounds;
- Altering Reactions;
- · Removing reactions and metabolites;
- Searching for duplicates and comparison of two models;
- Changing the model objective;
- Changing the direction of reaction(s);
- Creating gene-reaction-associations ('GPRs');
- · Extracting a subnetwork;

#### **EQUIPMENT SETUP**

Start CobraToolbox

initCobraToolbox;

The COBRA Toolbox - 2017

Documentation:

http://opencobra.github.io/cobratoolbox

- > Checking if git is installed ... Done.
- > Checking if the repository is tracked using git ... Done.
- > Checking if curl is installed ... Done.
- > Checking if remote can be reached ... Done.
- > Initializing and updating submodules ... Done.
- > Adding all the files of The COBRA Toolbox ... Done.
- > Define CB map output... set to svg.
- > Retrieving models ... Done.
- > TranslateSBML is installed and working properly.
- > Configuring solver environment variables ...
  - [---\*] ILOG\_CPLEX\_PATH: C:\Program Files\IBM\ILOG\CPLEX Studio1263\cplex\matlab\x64 win64
  - [----] GUROBI PATH : --> set this path manually after installing the solver ( see instructions )
  - [---\*] TOMLAB PATH: C:\Program Files\tomlab\
  - [----] MOSEK PATH : --> set this path manually after installing the solver ( see instructions ) Done.
- > Checking available solvers and solver interfaces ... Done.
- > Setting default solvers ... Done.
- > Saving the MATLAB path ... Done.
  - The MATLAB path was saved in the default location.
- > Summary of available solvers and solver interfaces

Support	LP	MILP		QP	MIQP	NL	Р		
cplex_direct	full		0		0	0	0	-	
dqqMinos	full		0		-	-	-	-	
glpk	full		1		1	-	-	-	
gurobi	full		1		1	1	1	-	
ibm_cplex	full		0		0	0	-	-	
matlab	full		1		-	-	-	1	
mosek	full		0		0	0	-	-	
pdco	full		1		-	1	-	-	
quadMinos	full		0		-	-	-	0	
tomlab_cplex	full		1		1	1	1	-	
qpng	experimental		-		-	1	-	-	
tomlab_snopt	experimental		-		-	-	-	1	
gurobi_mex	legacy		0		0	0	0	-	
lindo_old	legacy		0		-	-	-	-	
lindo_legacy	legacy		0		-	-	-	-	
lp_solve	legacy		1		-	-	-	-	
opti	legacy		0		0	0	0	0	
Total	-		6		3	4	2	2	

- + Legend: = not applicable, 0 = solver not compatible or not installed, 1 = solver installed.
- > You can solve LP problems using: 'glpk' 'gurobi' 'matlab' 'pdco' 'tomlab\_cplex' 'lp\_solve'
  > You can solve MILP problems using: 'glpk' 'gurobi' 'tomlab\_cplex'
- > You can solve QP problems using: 'gurobi' 'pdco' 'tomlab\_cplex' 'qpng'
- > You can solve MIQP problems using: 'gurobi' 'tomlab\_cplex'
- > You can solve NLP problems using: 'matlab' 'tomlab snopt'
- > Checking for available updates ...
- --> You cannot update your fork using updateCobraToolbox(). [f4a806 @ TutorialReview-ModelManipulation Please use the MATLAB.devTools (https://github.com/opencobra/MATLAB.devTools).

#### Generate a network

A constraint-based metabolic model contains the stoichiometric matrix with reactions and metabolites [1].

S is a stoichiometric representation of metabolic networks corresponding to the reactions in the biochemical pathway. In each column of the S is a biochemical reaction (n) and in each row is a precise metabolite (m). There is a stoichiometric coefficient of zero, which means that metabolite does not participate in that distinct reaction. The coefficient also can be positive when the appropriate metabolite is produced, or negative for every metabolite depleted [1].

```
GLCt1r glc_D[e] <=> glc_D[c]

HEX1 glc_D[c] + atp[c] -> h[c] + adp[c] + g6p[c]

PGI g6p[c] <=> f6p[c]

PFK atp[c] + f6p[c] -> h[c] + adp[c] + fdp[c]

FBP fdp[c] + h2o[c] -> f6p[c] + pi[c]

FBA fdp[c] <=> g3p[c] + dhap[c]

TPI dhap[c] <=> g3p[c]
```

We can now have a look at the different model fields created. The stoichiometry is stored in the S field of the model, which was described above. Since this is commonly a sparse matrix (i.e. it does contain a lot of zeros) to display it, it is useful to use the full representation:

```
full(model.S)
```

```
ans =
     - 1
             0
                    0
                            0
                                   0
                                                  0
     1
            - 1
                    0
                            0
                                   0
                                           0
                                                  0
     0
            - 1
                    0
                           - 1
                                   0
                                           0
                                                  0
     0
            1
                    0
                           1
                                   0
                                                  0
     0
             1
                    0
                            1
                                   0
                                           0
                                                  0
     0
             1
                    - 1
                            0
                                   0
                                           0
                                                  0
     0
             0
                    1
                           - 1
                                   1
                                           0
                                                  0
     0
             0
                    0
                           1
                                                  0
                                  - 1
                                          - 1
     0
             0
                    0
                            0
                                  - 1
                                          0
                                                  0
                                   1
     0
             0
                    0
                            0
                                           0
                                                  0
```

Some descriptive fields always present are model.mets and model.rxns which represent the metabolites and the reactions respectively.

```
model.mets
ans =
     'glc D[e]'
     'glc_D[c]'
     'atp[c]'
     'h[c]'
     'adp[c]'
     'g6p[c]'
     'f6p[c]'
     'fdp[c]'
     'h2o[c]'
     'pi[c]'
     'g3p[c]'
     'dhap[c]'
model.rxns
ans =
     'GLCt1r'
     'HEX1'
     'PGI'
     'PFK'
     'FBP'
     'FBA'
     'TPI'
```

Fields in the COBRA model are commonly column vectors, which can be an important detail when writing functions manipulating these fields.

There are a few more fields present in each COBRA model:

model.lb, indicating the lower bounds of each reaction, and model.ub indicating the upper bound of a reaction.

```
% this displays an array with reaction names and flux bounds.
[{'Reaction ID', 'Lower Bound', 'Upper Bound'};...
 model.rxns, num2cell(model.lb), num2cell(model.ub)]
ans =
    'Reaction ID'
                    'Lower Bound'
                                    'Upper Bound'
    'GLCt1r'
                    [ -20]
                                              201
                                    [
    'HEX1'
                             0]
                                              20]
                    [
                                    [
    'PGI'
                             -20]
                                              20]
                                    [
    'PFK'
                    [
                              0]
                                    [
                                              20]
    'FBP'
                                              20]
                    [
                              0]
                                    [
    'FBA'
                    [
                             -20]
                                    [
                                              20]
    'TPI'
                             -20]
                                    [
                                              20]
```

```
\$ This is a convenience function which does pretty much the same as the line above. 
  printFluxBounds (model); \\
```

```
Reaction ID Lower Bound Upper Bound
GLCt1r -20.000 20.000
```

```
HEX1
              0.000
                            20.000
            -20.000
 PGI
                            20,000
 PFK
             0.000
                            20,000
 FBP
             0.000
                            20.000
 FBA
            -20.000
                            20.000
 TPI
            -20.000
                            20.000
```

Before we start to modify the model, it might be useful to store some of the current properties of the model

```
mets_length = length(model.mets);
rxns_length = length(model.rxns);
```

# Creating, adding and handling reactions

If we want to add a reaction to the model or modify an existing reaction we are using function addReaction.

We will add some more reactions from glycolysis.

· The formula approach

```
model = addReaction(model, 'GAPDH',...
    'reactionFormula', 'g3p[c] + nad[c] + 2 pi[c] -> nadh[c] + h[c] + 13bpg[c]');

GAPDH 2 pi[c] + g3p[c] + nad[c] -> h[c] + nadh[c] + 13bpg[c]

model = addReaction(model, 'PGK',...
    'reactionFormula', '13bpg[c] + adp[c] -> atp[c] + 3pg[c]');

PGK adp[c] + 13bpg[c] -> atp[c] + 3pg[c]

model = addReaction(model, 'PGM', 'reactionFormula', '3pg[c] <=> 2pg[c]' );

PGM 3pg[c] <=> 2pg[c]
```

Display of stoichiometric matrix after adding reactions. Note the enlarge link when you move your mouse over the output to display the full matrix:

```
full(model.S)
ans =
    - 1
     1
          -1
                 0
                      0
                            0
     0
                 0
                                                    1
                                                          0
          - 1
                      - 1
                            0
                                  0
                                        0
     0
          1
                0
                      1
                            0
                                  0
                                        0
                                              1
                                                    0
                                                          0
     0
          1
                0
                     1
                            0
                                 0
                                        0
                                              0
                                                   - 1
                                                          0
     0
          1
                - 1
                      0
                            0
                                 0
                                        0
                                              0
                                                    0
                                                          0
     0
          0
                1
                                                          0
                      - 1
                            1
                                  0
                                        0
                                              0
                                                    0
     0
          0
                0
                      1
                                                    0
                                                          0
                            - 1
                                 - 1
                                        0
                                              0
     0
           0
                 0
                      0
                                  0
                                        0
                                              0
                                                    0
                                                          0
                            - 1
     0
           0
                 0
                      0
                                        0
                                                    0
                                                          0
                            1
                                  0
                                             -2
```

```
% rows(for nadh, nad, 13bpg, 2pg and 3pg metabolites)
```

If you want to search reactions sequence in the model and change the order of the selected reaction, use the following functions.

```
rxnID = findRxnIDs(model, 'GAPDH')

rxnID = 8

model = moveRxn(model, rxnID, 1);
rxnID = findRxnIDs(model, 'GAPDH')

rxnID = 1
```

While moving reactions does not modify the structure as such it can help in keeping a model tidy.

The list approach

The addReaction function has ability to recognize duplicate reactions when an order of metabolites and an abbreviation of the reaction are different.

```
model = addReaction(model, 'GAPDH2',...
    'metaboliteList', {'g3p[c]', 'nad[c]', 'pi[c]', '13bpg[c]', 'nadh[c]', 'h[c]' },...
    'stoichCoeffList', [-1; -1; -2; 1; 1; 1], 'reversible', false);
```

Since the second call should not have added anything we will check if the number of the reaction increased by the three reactions we added (and not by the one duplicated) and the number of metabolites was incremented by five (13bpg, nad, nadh, 23bpg, and 2pg).

```
assert(length(model.rxns) == rxns_length + 3);
assert(length(model.mets) == mets_length + 5);
```

## Adding Exchange, Sink and Demand reactions

The specific type of reactions in the constraint-based models are reactions that are using and recycling accumulated metabolites or producing required metabolites in the model.

- 1. *Exchange reactions* Reactions added to the model to move metabolites across the created *in silico* compartments. Those compartments represent intra- and intercellular membranes.
- 2. *Sink reactions* The metabolites, produced in reactions that are outside of an ambit of the system or in unknown reactions, are supplied to the network with reversible sink reactions.
- 3. *Demand reactions* Irreversible reactions added to the model to consume metabolites that are deposited in the system.

There are two ways to implement that kind of reactions:

1. Use addReaction with the documented function call:

```
model = addReaction(model, 'EX_glc_D[e]', 'metaboliteList', {'glc_D[e]'}, ...
```

```
'stoichCoeffList', [-1]);
```

```
EX_glc_D[e] glc_D[e] <=>
```

In the bigger networks we can find our exchange reactions with the following functions:

```
% determines whether a reaction is a general exchange reaction (selExc) and % whether its an uptake reaction (selUpt).
[selExc, selUpt] = findExcRxns(model, 0, 1)
```

```
selExc =
   0
   0
   0
   0
   0
   0
   0
   0
   0
   0
   1
selUpt =
   0
   0
   0
   0
   0
   0
   0
   0
   0
   0
   0
```

2. Use a utility function to create a particular reaction type: addExchangeRxn, addSinkReactions, addDemandReaction.

```
model = addExchangeRxn(model, {'glc D[e]', 'glc D[c]'})
EX_glc_D[e] glc_D[e]
EX glc D[c] glc D[c]
model =
          rxns: {12×1 cell}
             S: [17×12 double]
            lb: [12×1 double]
            ub: [12×1 double]
             c: [12×1 double]
          mets: {17×1 cell}
             b: [17×1 double]
         rules: {12×1 cell}
         genes: {0×1 cell}
        osense: -1
        csense: [17×1 char]
    rxnGeneMat: [12×0 double]
      rxnNames: {12×1 cell}
    subSystems: {12×1 cell}
      metNames: {17×1 cell}
       grRules: {12×1 cell}
```

```
model = addSinkReactions(model, {'13bpg[c]', 'nad[c]'})
sink 13bpg[c] 13bpg[c] <=>
sink nad[c] nad[c] <=>
model =
          rxns: {14×1 cell}
             S: [17×14 double]
            lb: [14×1 double]
            ub: [14×1 double]
             c: [14×1 double]
          mets: {17×1 cell}
             b: [17×1 double]
         rules: {14×1 cell}
         genes: {0×1 cell}
        osense: -1
        csense: [17×1 char]
    rxnGeneMat: [14×0 double]
      rxnNames: {14×1 cell}
    subSystems: {14×1 cell}
      metNames: {17×1 cell}
       grRules: {14×1 cell}
```

```
model = addDemandReaction(model, {'dhap[c]', 'q3p[c]'})
DM dhap[c] dhap[c] ->
DM_g3p[c] g3p[c] \rightarrow
model =
          rxns: {16×1 cell}
             S: [17×16 double]
            lb: [16×1 double]
            ub: [16×1 double]
             c: [16×1 double]
          mets: {17×1 cell}
             b: [17×1 double]
         rules: {16×1 cell}
         genes: {0×1 cell}
        osense: -1
        csense: [17×1 char]
    rxnGeneMat: [16×0 double]
      rxnNames: {16×1 cell}
    subSystems: {16×1 cell}
      metNames: {17×1 cell}
       grRules: {16×1 cell}
```

## Setting ratio between the reactions and changing reactions boundary

It is important to emphasise that previous knowledge base information should be taken into account. Most of them could disrupt future analysis of the model.

For instance, if it is familiar that flux through one reaction is *X* times the flux through another reaction, it is recommended to specify that in your model.

```
E.g. 1 v \text{ EX\_glc\_}D[c] = 2 v \text{ EX\_glc\_}D[e]
```

```
model = addRatioReaction (model, {'EX_glc_D[c]', 'EX_glc_D[e]'}, [1; 2]);
```

## **Altering Reaction bounds**

In order to respect the transport and exchange potential of a particular metabolite, or to resemble the different conditions in the model, we frequently need to set appropriate limits of the reactions.

```
model = changeRxnBounds(model, 'EX_glc_D[e]', -18.5, 'l');
```

# **Modifying Reactions**

The addReaction function also is a good choice when modifying reactions. By supplying a new stoichiometry, the old will be overwritten. For example, further up, we added a wrong stoichiometry for the GAP-Dehydrogenase with a phosphate coefficient of 2 (easily visible by printing the reaction).

```
printRxnFormula(model, 'rxnAbbrList', 'GAPDH');

GAPDH 2 pi[c] + g3p[c] + nad[c] -> h[c] + nadh[c] + 13bpg[c]
```

We can correct this by simply calling addReaction again with the corrected stoichiometry. In essence, parts which are not provided are taken from the old reaction, and only the new ones overwrite the existing data.

```
model = addReaction(model, 'GAPDH',...
    'metaboliteList', {'g3p[c]', 'nad[c]', 'pi[c]', '13bpg[c]', 'nadh[c]', 'h[c]' },...
    'stoichCoeffList', [-1; -1; 1; 1; 1]);

GAPDH pi[c] + g3p[c] + nad[c] -> h[c] + nadh[c] + 13bpg[c]
```

We might also want to add a gene rule to the reaction. This can be done using

```
model = changeGeneAssociation(model, 'GAPDH', 'G1 and G2');

New gene G1 added to model
New gene G2 added to model

printRxnFormula(model, 'rxnAbbrList', {'GAPDH'}, 'gprFlag', true);

GAPDH pi[c] + g3p[c] + nad[c] -> h[c] + nadh[c] + 13bpg[c] G1 and G2
```

Another option to achieve this is to use addReaction and the geneRule parameter.

```
model = addReaction(model, 'PGK', 'geneRule', 'G2 or G3', 'printLevel', 0);

New gene G3 added to model

printRxnFormula(model, 'gprFlag', true);

GAPDH pi[c] + g3p[c] + nad[c] -> h[c] + nadh[c] + 13bpg[c] G1 and G2
```

```
GLCt1r glc_D[e] <=> glc_D[c]
HEX1 glc_D[c] + atp[c] -> h[c] + adp[c] + g6p[c]
PGI g6p[c] <=> f6p[c]
PFK atp[c] + f6p[c] -> h[c] + adp[c] + fdp[c]
FBP fdp[c] + h2o[c] -> f6p[c] + pi[c]
FBA fdp[c] <=> g3p[c] + dhap[c]
TPI dhap[c] <=> g3p[c]
PGK adp[c] + 13bpg[c] -> atp[c] + 3pg[c] G2 or G3
PGM 3pg[c] <=> 2pg[c]
EX_glc_D[e] glc_D[e] <=> 2 Ratio_EX_glc_D[c]_EX_glc_D[e]
EX_glc_D[c] glc_D[c] + Ratio_EX_glc_D[c]_EX_glc_D[e] <=> sink_13bpg[c] 13bpg[c] <=> sink_nad[c] nad[c] <=> DM_dhap[c] dhap[c] -> DM_g3p[c] g3p[c] ->
```

#### Remove reactions and metabolites

In order to remove reactions from the model, the following function can been used:

#### Remove metabolites

```
model = removeMetabolites(model, {'3pg[c]', '2pg[c]'}, false);
```

For instance, in the previous code, many metabolites from 'GAPDH' were deleted, but the reaction is still present in the model (since there are more metabolites left). The false indicates, that empty reactions should not be removed.

To delete unused metabolites that participate in no reaction and reactions that have no participating metabolites (all zero rows and columns in the *S* matrix), the following function can be used:

```
model = removeTrivialStoichiometry(model)
```

```
model =
          rxns: {9×1 cell}
             S: [15×9 double]
            lb: [9×1 double]
            ub: [9×1 double]
             c: [9×1 double]
          mets: {15×1 cell}
             b: [15×1 double]
         rules: {9×1 cell}
         genes: {3×1 cell}
        osense: -1
        csense: [15×1 char]
    rxnGeneMat: [9×3 double]
      rxnNames: {9×1 cell}
    subSystems: {9×1 cell}
      metNames: {15×1 cell}
       grRules: {9×1 cell}
          note: 'EX glc D[c] andEX glc D[e]are set to have a ratio of1:2.'
```

## Search for duplicates and comparison of two models

assert(rxns length + 2 == length(model.rxns));

Since genome-scale metabolic models are expanding every day [2], the need for comparison of them is also spreading.

The elementary functions for the model manipulation, besides main actions, simultaneously perform the structural analysis and comparison (e.g. addReaction). Likewise, there are additional functions that are only dealing with analyzing similarities and differences within and between the models.

• Checking for reaction duplicates by reaction abbreviation, by using method 'S' that will not detect reverse reactions and method 'FR' that will neglect reactions direction:

```
[model, removedRxn, rxnRelationship] = checkDuplicateRxn(model, 'S', 1, 1);
  Checking for reaction duplicates by stoichiometry ...
   no duplicates found.
Adding duplicate reaction to the model:
  model = addReaction(model, 'GLCt1r duplicate reverse',...
                        'metaboliteList', {'glc_D[e]', 'glc_D[c]'},...
                        'stoichCoeffList', [1 -1], 'lowerBound', 0, ...
'upperBound', 20, 'checkDuplicate', 0);
  GLCt1r duplicate reverse glc_D[c] -> glc_D[e]
  fprintf('>> Detecting duplicates using S method\n');
  >> Detecting duplicates using S method
  method = 'S';
 % will not be removed as does not detect a reverse reaction
  [model, removedRxn, rxnRelationship] = checkDuplicateRxn(model, method, 1, 1);
  Checking for reaction duplicates by stoichiometry ...
   no duplicates found.
  assert(rxns length + 3 == length(model.rxns));
 % The reaction length has been reevaluated
  fprintf('>> Detecting duplicates with using FR method\n');
  >> Detecting duplicates with using FR method
  method = 'FR';
  % will be removed as detects a reverse reaction
  [model, removedRxn, rxnRelationship] = checkDuplicateRxn(model, method, 1, 1);
  Checking for reaction duplicates by stoichiometry (up to orientation) ...
  Keep: GLCt1r glc_D[e] <=> glc_D[c]
Duplicate: GLCt1r_duplicate_reverse glc_D[c] -> glc_D[e]
```

• Function checkCobraModelUnique marks the reactions and metabolites that are not unique in the model.

# model = checkCobraModelUnique(model, false)

```
model =
         rxns: {9×1 cell}
             S: [15×9 double]
            lb: [9×1 double]
            ub: [9×1 double]
             c: [9×1 double]
          mets: {15×1 cell}
             b: [15×1 double]
         rules: {9×1 cell}
         genes: {3×1 cell}
        osense: -1
        csense: [15×1 char]
    rxnGeneMat: [9×3 double]
      rxnNames: {9×1 cell}
    subSystems: {9×1 cell}
     metNames: {15×1 cell}
      grRules: {9×1 cell}
          note: 'EX_glc_D[c] andEX_glc_D[e]are set to have a ratio of1:2.'
```

# Changing the model's objective

Simulating different conditions in the model is often necessary for a favor of performing calculations that investigate a specific objective. One of the elementary objectives is optimal growth [3]. The model can be modified to get different conditions with changing the model objective:

```
modelNew = changeObjective(model, 'GLCt1r', 0.5);
% multiple reactions, default coefficient (1)
modelNew = changeObjective(model, {'PGI'; 'PFK'; 'FBP'});
```

#### The direction of reactions

For some purposes, it is important to only have irreversible reactions in a model, i.e. only allowing positive flux in all reactions. This can be important if e.g. absolute flux values are of interest and negative flux would reduce an objective while it should actually increase it. The COBRA Toolbox offers functionality to change a model to an irreversible format, by splitting all reversible reactions and adjusting the respective lower and upper bounds, such that the model capacities stay the same.

Let us see, how the glycolysis model currently looks:

```
printRxnFormula(model);
```

```
\begin{split} & \mathsf{GAPDH} \ \mathsf{pi[c]} + \mathsf{g3p[c]} + \mathsf{nad[c]} \ -> \mathsf{h[c]} + \mathsf{nadh[c]} + \mathsf{13bpg[c]} \\ & \mathsf{GLCt1r} \ \mathsf{glc\_D[e]} \ <=> \ \mathsf{glc\_D[c]} \\ & \mathsf{HEX1} \ \mathsf{glc\_D[c]} + \mathsf{atp[c]} \ -> \mathsf{h[c]} + \mathsf{adp[c]} + \mathsf{g6p[c]} \\ & \mathsf{PGI} \ \mathsf{g6p[c]} \ <=> \ \mathsf{f6p[c]} \\ & \mathsf{PFK} \ \mathsf{atp[c]} + \mathsf{f6p[c]} \ -> \mathsf{h[c]} + \mathsf{adp[c]} + \mathsf{fdp[c]} \\ & \mathsf{FBP} \ \mathsf{fdp[c]} + \mathsf{h2o[c]} \ -> \ \mathsf{f6p[c]} + \mathsf{pi[c]} \\ & \mathsf{FBA} \ \mathsf{fdp[c]} \ <=> \ \mathsf{g3p[c]} + \mathsf{dhap[c]} \end{split}
```

```
TPI dhap[c] <=> g3p[c]
PGK adp[c] + 13bpg[c] -> atp[c]
```

To convert a model to an irreversible model use the following command:

```
[modelIrrev, matchRev, rev2irrev, irrev2rev] = convertToIrreversible(model);
```

Compare the irreversible model with the original model:

```
printRxnFormula(modelIrrev);
```

```
GAPDH pi[c] + g3p[c] + nad[c] -> h[c] + nadh[c] + 13bpg[c]
GLCt1r_f glc_D[e] -> glc_D[c]
HEX1 glc_D[c] + atp[c] -> h[c] + adp[c] + g6p[c]
PGI_f g6p[c] -> f6p[c]
PFK atp[c] + f6p[c] -> h[c] + adp[c] + fdp[c]
FBP fdp[c] + h2o[c] -> f6p[c] + pi[c]
FBA_f fdp[c] -> g3p[c] + dhap[c]
TPI_f dhap[c] -> g3p[c]
PGK adp[c] + 13bpg[c] -> atp[c]
GLCt1r_b glc_D[c] -> glc_D[e]
PGI_b f6p[c] -> g6p[c]
FBA_b g3p[c] + dhap[c] -> fdp[c]
TPI_b g3p[c] -> dhap[c]
```

You will notice, that there are more reactions in this model and that all reactions which have a lower bound < 0 are split in two.

There is also a function to convert an irreversible model to a reversible model:

```
modelRev = convertToReversible(modelIrrev);
```

If we now compare the reactions of this model with those from the original model, they should look the same.

```
printRxnFormula(modelRev);
```

```
 \begin{array}{lll} \text{GAPDH pi[c]} + \text{g3p[c]} + \text{nad[c]} & -> \text{h[c]} + \text{nadh[c]} + 13 \text{bpg[c]} \\ \text{GLCt1r glc\_D[e]} & <=> \text{glc\_D[c]} \\ \text{HEX1 glc\_D[c]} + \text{atp[c]} & -> \text{h[c]} + \text{adp[c]} + \text{g6p[c]} \\ \text{PGI g6p[c]} & <=> \text{f6p[c]} \\ \text{PFK atp[c]} + \text{f6p[c]} & -> \text{h[c]} + \text{adp[c]} + \text{fdp[c]} \\ \text{FBP fdp[c]} + \text{h2o[c]} & -> \text{f6p[c]} + \text{pi[c]} \\ \text{FBA fdp[c]} & <=> \text{g3p[c]} + \text{dhap[c]} \\ \text{TPI dhap[c]} & <=> \text{g3p[c]} \\ \text{PGK adp[c]} + 13 \text{bpg[c]} & -> \text{atp[c]} \\ \end{array}
```

Create gene-reaction-associations ('GPRs') from scratch.

Assign the GPR '(G1) or (G2)' to the reaction HEX1

```
model = changeGeneAssociation(model, 'HEX1', '(G1) or (G2)');
```

Check that there are no empy columns left.

```
find(sum(model.rxnGeneMat, 1) == 0)
```

```
ans =
  1×0 empty double row vector
```

## Replace a GPR with a new one.

Here, we will search for all instances of a specific GPR ('G1 or G2') and replace it with a new one ('G1 or G4').

Define the old and the new GPRs.

```
GPRsReplace = {'G1 or G2' 'G1 or G4'};
```

Find all reactions that have the old GPR

```
for i = 1 : size(GPRsReplace, 1)
  oldGPRrxns = find(strcmp(model.grRules, GPRsReplace{i, 1}));
  for j = 1:length(oldGPRrxns)
      model = changeGeneAssociation(model, model.rxns{oldGPRrxns(j)}, GPRsReplace{i, 2});
  end
end
```

# Remove unused genes

Let's assume that the reaction PGK has to be removed from the model

```
model = removeRxns(model, 'PGK');
```

The model now contains genes that do not participate in any GPR

```
find(sum(model.rxnGeneMat, 1) == 0)
ans = 3
```

We remove unused genes by re-assigning the model's GPR rules, which updates the reaction-genematrix and gene list.

Store GPR list in a new variable

```
storeGPR = model.grRules;
```

Erase model's gene list and reaction-gene-matrix

```
model.rxnGeneMat = [];
model.genes = [];
```

Re-assign GPR rules to model

```
for i = 1 : length(model.rxns)
    model = changeGeneAssociation(model, model.rxns{i}, storeGPR{i});
end
```

```
New gene G2 added to model
```

Check that there are no unused genes left in the model.

```
find(sum(model.rxnGeneMat, 1) == 0)

ans =
  1×0 empty double row vector
```

#### Remove common issues with GPR definitions and reaction abbreviations

Remove issues with quotation marks in the GPR definitions.

```
model.grRules = strrep(model.grRules, '''', '');
```

Remove spaces from reaction abbreviations.

```
model.rxns = strrep(model.rxns, ' ', '');
```

Remove unneccessary brackets from the GPR associations.

```
for i = 1 : length(model.grRules)
   if isempty(strfind(model.grRules{i}, 'and')) && isempty(strfind(model.grRules{i}, 'or'))%
        model.grRules{i} = regexprep(model.grRules{i}, '[\(\\)]', '');
   end
end
```

### **Extract subnetwork**

Extract a subnetwork from the model consisting of the reactions HEX1, PGI, FBP, and FBA. The function will remove unused metabolites.

```
rxnList = {'HEX1'; 'PGI'; 'FBP'; 'FBA'}

rxnList =
    'HEX1'
    'PGI'
    'FBP'
    'FBA'
```

```
subModel = extractSubNetwork(model, rxnList)
```

```
subModel =
    rxns: {4×1 cell}
        S: [11×4 double]
        lb: [4×1 double]
        ub: [4×1 double]
        c: [4×1 double]
        mets: {11×1 cell}
        b: [11×1 double]
        rules: {4×1 cell}
        genes: {2×1 cell}
        osense: -1
        csense: [11×1 char]
```

rxnGeneMat: [4×2 double]
 rxnNames: {4×1 cell}
subSystems: {4×1 cell}
metNames: {11×1 cell}
 grRules: {4×1 cell}
 note: 'EX\_glc\_D[c] andEX\_glc\_D[e]are set to have a ratio of1:2.'

#### **REFERENCES**

[1] Orth, J. D., Thiele I., and Palsson, B. Ø. (2010). What is flux balance analysis? *Nat. Biotechnol.*, 28(3), 245–248.

[2] Feist, A. M., Palsson, B. (2008). The growing scope of applications of genome-scale metabolic reconstructions: the case of *E. coli. Nature Biotechnology*, *26*(6), 659–667.

[3] Feist, A. M., Palsson, B. O. (2010). The Biomass Objective Function. *Current Opinion in Microbiology*, *13*(3), 344–349.