Flux Variability analysis (FVA)

All COBRA predictions are derived from optimisation problems, typically formulated in the form

$$\begin{aligned} \min_{v \in R^n} \quad & \psi(v) \\ \text{s.t.} \quad & Sv = b, \\ & Cv \leq d, \\ & l \leq v \leq u, \end{aligned}$$

where $v \in R^n$ represents the rate of each biochemical reaction, and where a $b_i < 0$, or $b_i > 0$, represents some fixed output, or input, of the ith molecular species. $S \in \Re^{m \times n}$ is a stoichiometric matrix for m molecular species and n reactions, and b is a vector of known metabolic exchanges. The output of FBA is a particular flux distribution, v , which maximises or minimises the objective function and stands between upper and lower bounds, u and l, respectively.

The Flux variability analysis (FVA) is a computational tool that is useful for quantifying excess fluxes by identifying the set of metabolic phenotypes that result in an equivalent maximal (or minimal) objective function. This analysis is essential for identifying biochemical pathways that can potentially generate the same phenotype. It computes the minimal and maximal fluxes of each reaction for a fixed objective function [1].

The FVA is used for examining changes in the distribution of metabolic fluxes under several altered conditions [1]. It solves two optimisation problems for each flux v_i of interest:

$$\begin{aligned} \max_v / \min_v \quad & v_i \\ \text{s.t.} \quad & Sv = b, \\ & w^T v \geq ‡ Z_0, \\ & l \leq v \leq u, \end{aligned}$$

where w is a biological objective, such as ATP production, and $Z_0 = w^T v_0$ ia an optimal solution to LP, γ is a parameter which controls whether the analysis is done w.r.t. suboptimal network states $(0 \leq \gamma \leq 1)$ or to the optimal state $(\gamma = 1)$.

EQUIPMENT SETUP

If necessary, initialize the cobra toolbox:

```
initCobraToolbox
```

```
 _____   _____   _____   _____     _____
/  ___| /  _  \ |  _  \ |  _  \   / ___ \
| |     | | | | | |_| | | |_| |   | |___| |
| |     | | | | |  _  { |  _  /   |  ___  |
| |___  | |_| | | |_| | | | \ \   | |   | |
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COnstraint-Based Reconstruction and Analysis
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 :  --> set this path manually after installing the solver ( see instructions
  - [*---] GUROBI_PATH: /opt/gurobi702/linux64/matlab
  - [----] TOMLAB_PATH :  --> set this path manually after installing the solver ( see instructions )
  - [----] 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 as ~/pathdef.m.

> Summary of available solvers and solver interfaces

  Support     LP   MILP     QP   MIQP    NLP
-----------------------------------------------------------------------
cplex_direct  full              0      0      0      0      -
dqqMinos      full              1      -      -      -      -
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              1      -      -      -      1
tomlab_cplex  full              0      0      0      0      -
qpng          experimental      -      -      1      -      -
tomlab_snopt  experimental      -      -      -      -      0
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         -                 7      2      3      1      2

+ Legend: - = not applicable, 0 = solver not compatible or not installed, 1 = solver installed.


> You can solve LP problems using: 'dqqMinos' - 'glpk' - 'gurobi' - 'matlab' - 'pdco' - 'quadMinos' - '
> You can solve MILP problems using: 'glpk' - 'gurobi'
> You can solve QP problems using: 'gurobi' - 'pdco' - 'qpng'
> You can solve MIQP problems using: 'gurobi'
> You can solve NLP problems using: 'matlab' - 'quadMinos'

> Checking for available updates ...
> The COBRA Toolbox is up-to-date.
```

For solving linear programming problems in FBA and FVA analysis, certain solvers are required:

```
changeCobraSolver ('gurobi', 'all', 1);
```

```
> Gurobi interface added to MATLAB path.
> Solver for LPproblems has been set to gurobi.

> Gurobi interface added to MATLAB path.
> Solver for MILPproblems has been set to gurobi.

> Gurobi interface added to MATLAB path.
```

```
> Solver for QPproblems has been set to gurobi.

> Gurobi interface added to MATLAB path.
> Solver for MIQPproblems has been set to gurobi.
> Solver gurobi not supported for problems of type NLP. Currently used: matlab
```

The present tutorial can run with `'glpk'` package, which does not require additional installation and configuration. Although, for the analysis of large models is recommended to use the `'gurobi'` package. For detail information, refer to the solver installation guide: https://github.com/opencobra/cobratoolbox/blob/master/docs/source/installation/solvers.md

PROCEDURE

Before proceeding with the simulations, the path for the model needs to be set up:

```
pathModel = '~/work/sbgCloud/data/models/published/thiele_candidate_2005/';
filename = 'cardiac_mit_glcuptake_atpmax.mat';
load([pathModel, filename])
model = modelCardioMito;
clear modelCardioMito
```

In this tutorial, the provided model is a human cardiac mitochondrial model [2].

In this example, we are analysing the variability of several reactions from the human cardiac mitochondrial model in the aerobic and anaerobic state.

First, we will close the boundaries from uptake reactions and set constraints for two different biological states:

```
modelfva1 = model;
exchanges = { 'EX_12dgr_m(e)'
    'EX_co2(e)'
    'EX_coa(e)'
    'EX_cys-L(e)'
    'EX_fe2(e)'
    'EX_glc(e)'
    'EX_gly(e)'
    'EX_glyc(e)'
    'EX_glyc3p(e)'
    'EX_h(e)'
    'EX_h2o(e)'
    'EX_o2(e)'
    'EX_pi(e)'
    'EX_ps_m(e)'
    'sink_cdpchol(c)'
    'sink_cmp(c)'};
for i = 1:length (exchanges)
    modelfva1 = changeRxnBounds (modelfva1, exchanges{i}, 0, 'l');
end
 modelfva1 = changeRxnBounds (modelfva1, 'EX_glc(e)', -20, 'l');
 %modelfva2 represents aerobic condition
 modelfva2 = modelfva1;
  modelfva2 = changeRxnBounds (modelfva2, 'EX_o2(e)', -1000, 'l');
```

Running `fluxVariability()` on both models (`modelfva1`, `modelfva2`) will generate the minimum and maximum flux ranges of all the ractions in the network.

The code is as follows-

```
[minFlux1, maxFlux1, Vmin1, Vmax1] = fluxVariability(modelfva1, 0);
[minFlux2, maxFlux2, Vmin2, Vmax2] = fluxVariability(modelfva2, 0);
```

By fixing the `optPercentage` to zero, we specified no objective for the performed simulation.

Plotting several reactions from the results of the FVA and comparing them between the models:

```
index_max = find(maxFlux1 ~=0);
index_min = find(minFlux1 ~=0);
index = [index_max;index_min];
index = unique(index);
%obtain values in modelFva1
fav1_max = maxFlux1(index,1);
fav1_min = minFlux1(index,1);
%obtain values in modelFva2
fav2_max = maxFlux2(index,1);
fav2_min = minFlux2(index,1);

% find extreme values
i = find(fav1_max < 990);

%obtain new list of bounds and reactions
fav1_max = fav1_max(i,1);
fav1_min = fav1_min(i,1);
%obtain values in modelFva2
fav2_max = fav2_max(i,1);
fav2_min = fav2_min(i,1);

ymax1 = fav1_max;
ymin1 = fav1_min;
ymax2 = fav2_max;
ymin2 = fav2_min;

max =table(ymax1,ymax2)
```

```
max =
    ymax1    ymax2
    _____    _____

    40       638.2
     0           0
    40        61.8
    40          40
    40          40
    20          20
    40          40
    20          20
    20          20
     0           0
     0           0
    20          20
    20          20
     0           0
     0           0
```

```
    40          80
    20          20
```

```
min =table(ymin1,ymin2)
```

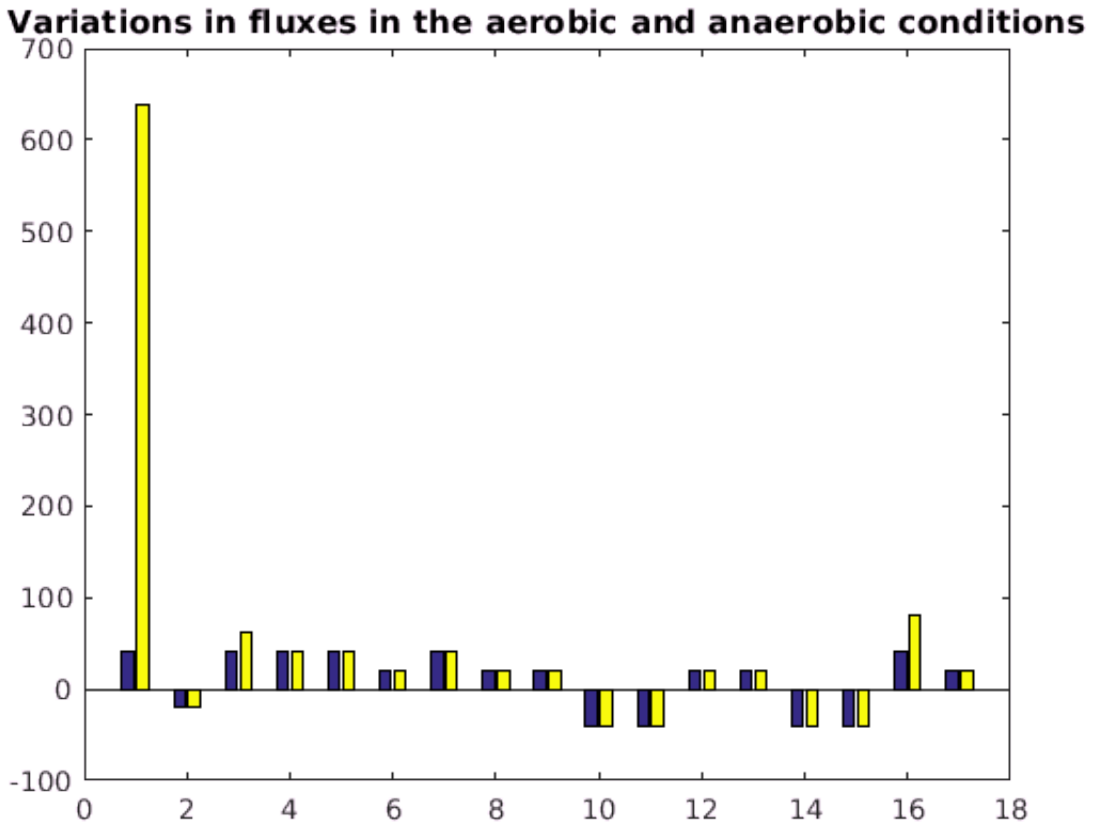
```
min =
    ymin1    ymin2
    _____    _____

      0        0
    -20      -20
      0        0
      0        0
      0        0
      0        0
      0        0
      0        0
      0        0
    -40      -40
    -40      -40
      0        0
      0        0
    -40      -40
    -40      -40
      0        0
      0        0
```

```
max = table2cell(max);
min = table2cell(min);

figure
plot1 = bar(cell2mat(max(1:end,:)));

hold on
plot2 = bar(cell2mat(min(1:end,:)));
title('Variations in fluxes in the aerobic and anaerobic conditions')
```



REFERENCES

[1] Gudmundsson S., and Thiele I., Computationally efficient flux variability analysis. *BMC Bioinformatics* 11:489 (2010).

[2] Thiele I., et al. Candidate Metabolic Network States in Human Mitochondria. Impact of diabetes, ischemia and diet. *J Bio Chem., 280*, 11683–11695 (2005).