# Sensitivity of a flux balance analysis solution with respect to input data

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## INTRODUCTION

Consider an FBA problem

$$\max c^{T} v$$
s.t.  $Sv = b$ 

$$l \le v \le u$$

The local sensitivity of the optimal objective value  $\mathcal{L}^{\star} = c^T v^{\star}$  with respect to a changes in the input data  $\{b, l, u\}$  is given by

$$\frac{\partial \mathcal{L}^{\star}}{\partial h} = y^{\star}$$

$$\frac{\partial \mathcal{L}^{\star}}{\partial l} = -w_l^{\star}$$

$$\frac{\partial \mathcal{L}^{\star}}{\partial u} = w_u^{\star}$$

where  $y^*$  is a vector of shadow prices and  $w = w_l - w_u$  is a vector of reduced costs. That is, a shadow price is the partial derivative of the optimal value of the objective function with respect to  $b_i$ . It indicates how much net production, or net consumption, of each metabolite increases (positive), or decreases (negative), the optimal value of the objective. The reduced costs,  $-w_l$  and  $w_u$  are the partial derivative of the optimal value of the objective function with respect to the lower and upper bounds on a reaction, respectively. They indicate how much relaxation, or tightening, of each bound increases, or decreases, the optimal objective, respectively. In the COBRA Toolbox, shadow prices and reduced costs are calculated by optimizeCbModel. When using the function

```
FBAsolution = optimizeCbModel(model,'max');
```

the shadow prices and reduced costs are given by FBAsolution.y and FBAsolution.w, respectively.

For a more complete theoretical description, see: cobratoolbox/tutorials/intro sensitivityAnalysis.pdf

### **MATERIALS - EQUIPMENT SETUP**

Please ensure that all the required dependencies (e.g., git and curl) of The COBRA Toolbox have been properly installed by following the installation guide here. Please ensure that the COBRA Toolbox has been initialised (tutorial\_initialize.mlx) and verify that the pre-packaged LP and QP solvers are functional (tutorial\_verify.mlx).

## **PROCEDURE**

## Load E. coli core model

The most direct way to load a model into The COBRA Toolbox is to use the readCbModel function. For example, to load a model from a MAT-file, you can simply use the filename (with or without file extension).

```
fileName = 'ecoli_core_model.mat';
if ~exist('modelOri','var')
modelOri = readCbModel(fileName);
end
%backward compatibility with primer requires relaxation of upper bound on
%ATPM
modelOri = changeRxnBounds(modelOri,'ATPM',1000,'u');
model = modelOri;
%setp the matlab e.coli metabolic map parameters
outputFormatOK = changeCbMapOutput('matlab');
map=readCbMap('ecoli_core_map');
options.zeroFluxWidth = 0.1;
options.rxnDirMultiplier = 10;
```

model 🔀 1x1 struct with 28 fields

Field ▲	Value	Size
<u>&gt;&gt;</u> S	72x95 sparse do	72×95
🚺 mets	72x1 cell	72x1
<mark>⊞</mark> b	72x1 double	72x1
🕩 csense	72x1 char	72x1
🚺 rxns	95x1 cell	95×1
<mark>⊞</mark> lb	95x1 double	95×1
<mark>⊞</mark> ub	95x1 double	95×1
<del>   </del> c	95x1 double	95×1
🕕 osenseStr	'max'	1x3
genes	137x1 cell	137x1
🚺 rules	95x1 cell	95x1
냂 metCharges	72x1 int32	72×1
🔱 metFormulas	72x1 cell	72x1
🔱 metNames	72x1 cell	72×1
metInChIString	72x1 cell	72×1
metKEGGID	72x1 cell	72×1
metChEBIID	72x1 cell	72×1
metPubChemID	72x1 cell	72x1
🚺 grRules	95x1 cell	95x1
rxnGeneMat	95x137 sparse d	95x137
rxnConfidence	95x1 double	95x1
🚺 rxnNames	95x1 cell	95x1
rxnNotes	95x1 cell	95x1
rxnECNumbers	95x1 cell	95x1
rxnReferences	95x1 cell	95x1

The meaning of each field in a standard model is defined in the standard COBRA model field definition.

In general, the following fields should always be present:

- S, the stoichiometric matrix
- mets, the identifiers of the metabolites
- **b**, Accumulation (positive) or depletion (negative) of the corresponding metabolites. 0 Indicates no concentration change.
- csense, indicator whether the b vector is a lower bound ('G'), upper bound ('L'), or hard constraint 'E' for the metabolites.
- rxns, the identifiers of the reactions
- **Ib**, the lower bounds of the reactions
- ub, the upper bounds of the reactions
- c, the linear objective
- genes, the list of genes in your model
- rules, the Gene-protein-reaction rules in a computer readable format present in your model.
- osenseStr, the objective sense either 'max' for maximisation or 'min' for minimisation

# **Sensitivity Analysis**

In the E. coli core model, when maximising ATP production, what is the shadow price of cytosolic protons?

Hint: FBAsolution.y

What is your biochemical interpretation of this change in objective in the current context?

Hint: printFluxVector, drawFlux

Perturb the model in such a way as to increase the optimal rate of ATP hydrolysis ('ATPM') by exactly one unit. How does this compare with the theoretical prediction?

Hint: change model.b

In the E. coli core model, when maximising ATP production, what is the reduced cost of glucose exchange?

Hint: FBAsolution.rcost

What is your biochemical interpretation of this?

Hint: use drawFlux with a perturbed optimal reaction rate vector

Perturb the model in such a way as to increase the optimal rate of ATP hydrolysis ('ATPM') by exactly 17.5 units. How does this compare with the theoretical prediction?

Hint: change model.lb

#### **TROUBLESHOOTING**

Note that, if an optimization problem is reformulated from a maximisation to a minimisation problem, then the signs of each of the dual variables is reversed.

#### **TIMING**

1 hr.

#### ANTICIPATED RESULTS

Understanding of how an optimal objective will change in response to changing the input data.

# **Acknowledgments**

Part of this tutorial was originally written by Jeff Orth and Ines Thiele for the publication "What is flux balance analysis?"

## **REFERENCES**

- 1. Orth. J., Thiele, I., Palsson, B.O., What is flux balance analysis? Nat Biotechnol. Mar; 28(3): 245–248 (2010).
- 2. Laurent Heirendt & Sylvain Arreckx, Thomas Pfau, Sebastian N. Mendoza, Anne Richelle, Almut Heinken, Hulda S. Haraldsdottir, Jacek Wachowiak, Sarah M. Keating, Vanja Vlasov, Stefania Magnusdottir, Chiam Yu Ng, German Preciat, Alise Zagare, Siu H.J. Chan, Maike K. Aurich, Catherine M. Clancy, Jennifer Modamio, John T. Sauls, Alberto Noronha, Aarash Bordbar, Benjamin Cousins, Diana C. El Assal, Luis V. Valcarcel, Inigo Apaolaza, Susan Ghaderi, Masoud Ahookhosh, Marouen Ben Guebila, Andrejs Kostromins, Nicolas Sompairac, Hoai M. Le, Ding Ma, Yuekai Sun, Lin Wang, James T. Yurkovich, Miguel A.P. Oliveira, Phan T. Vuong, Lemmer P. El Assal, Inna Kuperstein, Andrei Zinovyev, H. Scott Hinton, William A. Bryant, Francisco J. Aragon Artacho, Francisco J. Planes, Egils Stalidzans, Alejandro Maass, Santosh Vempala, Michael Hucka, Michael A. Saunders, Costas D. Maranas, Nathan E. Lewis, Thomas Sauter, Bernhard Ø. Palsson, Ines Thiele, Ronan M.T. Fleming, Creation and analysis of biochemical constraint-based models: the COBRA Toolbox v3.0, Nature Protocols, volume 14, pages 639–702, 2019 doi.org/10.1038/s41596-018-0098-2.