

Varying Parameters analysis

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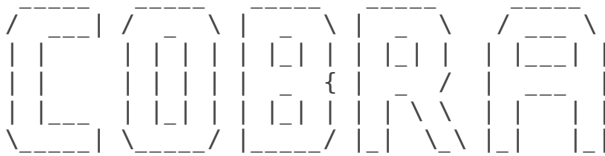
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In this tutorial, we show how computations are performed by varying one or two parameters over a fixed range of numerical values.

EQUIPMENT SETUP

If necessary, initialise the cobra toolbox:

```
initCobraToolbox;
```



COstraint-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	-
dqqMinos	full			1	-	-	-
glpk	full			1	1	-	-
gurobi	full			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 the 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/unpublished/Recon3D_models/';
filename = '2017_04_28_Recon3d.mat';
load([pathModel, filename])
model = modelRecon3model;
clear modelRecon3model
```

In this tutorial, the provided model is a generic model of the human cellular metabolism, Recon 3D [1]. Therefore, we assume, that the cellular objectives include energy production or optimisation of uptake rates and by-product secretion for various physiological functions of the human body.

TROUBLESHOOTING

If there are multiple energy sources available in the model; Specifying more constraints is necessary. If we do not do that, we will have additional carbon and oxygen energy sources available in the cell and the maximal ATP production.

To avoid this issue, all external carbon sources need to be closed.

```
%Closing the uptake of all energy and oxygen sources
idx=strmatch('Exchange/demand reaction', model.subSystems);
c=0;
for i=1:length(idx)
    if model.lb(idx(i))~=0
        c=c+1;
        uptakes{c}=model.rxns{idx(i)};
    end
end

modelalter = model;
modelalter = changeRxnBounds(modelalter, uptakes, 0, 'b');

% The alternative way to do that, in case you were using another large model,
% that does not contain defined Subsystem is
% to find uptake exchange reactions with following codes:
% [selExc, selUpt] = findExcRxns(model);
% uptakes = model.rxns(selUpt);
% Selecting from the exchange uptake reactions those
% which contain at least 1 carbon in the metabolites included in the reaction:
% subuptakeModel = extractSubNetwork(model, uptakes);
% hiCarbonRxns = findCarbonRxns(subuptakeModel,1);
% Closing the uptake of all the carbon sources
% modelalter = model;
% modelalter = changeRxnBounds(modelalter, hiCarbonRxns, 0, 'l');
```

Robustness analysis

Robustness analysis is applied to estimate and visualise how changes in the concentration of an environmental parameter (exchange rate) or internal reaction effect on the objective [2]. If we are interested in varying v_j between two values, i.e., v_{jmin} and v_{jmax} , we can solve l optimisation problems:

$$\begin{array}{ll} \max Z_k = c^T v & \\ \text{s.t.} & k = 1, \dots, l, \\ & Sv = 0, \\ \text{fixing} & v_j = v_{jmin} + \frac{(k-1)}{(l-1)} * (v_{jmax} - v_{jmin}) \\ \text{constraints} & v_{i,min} \leq v_i \leq v_{i,max}, i = 1, \dots, n, i \neq j \end{array}$$

The function `robustnessAnalysis()` is used for this analysis:

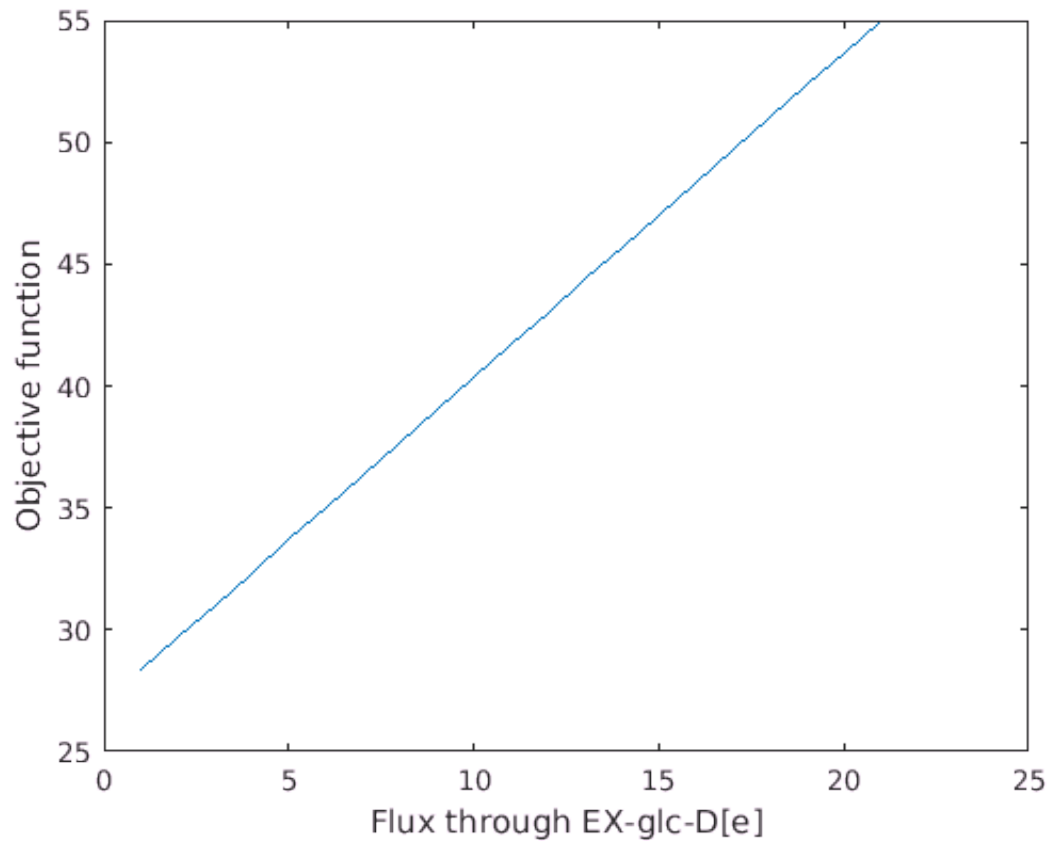
```
% [controlFlux, objFlux] = robustnessAnalysis(model, controlRxn, nPoints,...
%     plotResFlag, objRxn,objType)
```

where inputs are a COBRA model, a reaction that has been analysed and optional inputs:

```
% INPUTS
% model          COBRA model structure
% controlRxn     Reaction of interest whose value is to be controlled
%
% OPTIONAL INPUTS
% nPoints        Number of points to show on plot (Default = 20)
% plotResFlag    Plot results (Default true)
% objRxn         Objective reaction to be maximized
%                (Default = whatever is defined in model)
% objType        Maximize ('max') or minimize ('min') objective
%                (Default = 'max')
%
% OUTPUTS
% controlFlux    Flux values within the range of the maximum and minimum for
%                a reaction of interest
% objFlux        Optimal values of objective reaction at each control
%                reaction flux value
```

Here, we will investigate how robust the maximal ATP production of the network (i.e., the maximal flux through 'DM_atp_c_') is with respect to varying glucose uptake rates and fixed oxygen uptake.

```
modelrobust = modelalter;
modelrobust = changeRxnBounds(modelrobust, 'EX_o2[e]', -17, 'b');
AtpRates = zeros(21, 1);
for i = 0:20
    modelrobust = changeRxnBounds(modelrobust, 'EX_glc_D[e]', -i, 'b');
    modelrobust = changeObjective(modelrobust, 'DM_atp_c_');
    FBArobust = optimizeCbModel(modelrobust, 'max');
    AtpRates(i+1) = FBArobust.f;
end
plot (1:21, AtpRates)
xlabel('Flux through EX-glc-D[e]')
ylabel('Objective function')
```

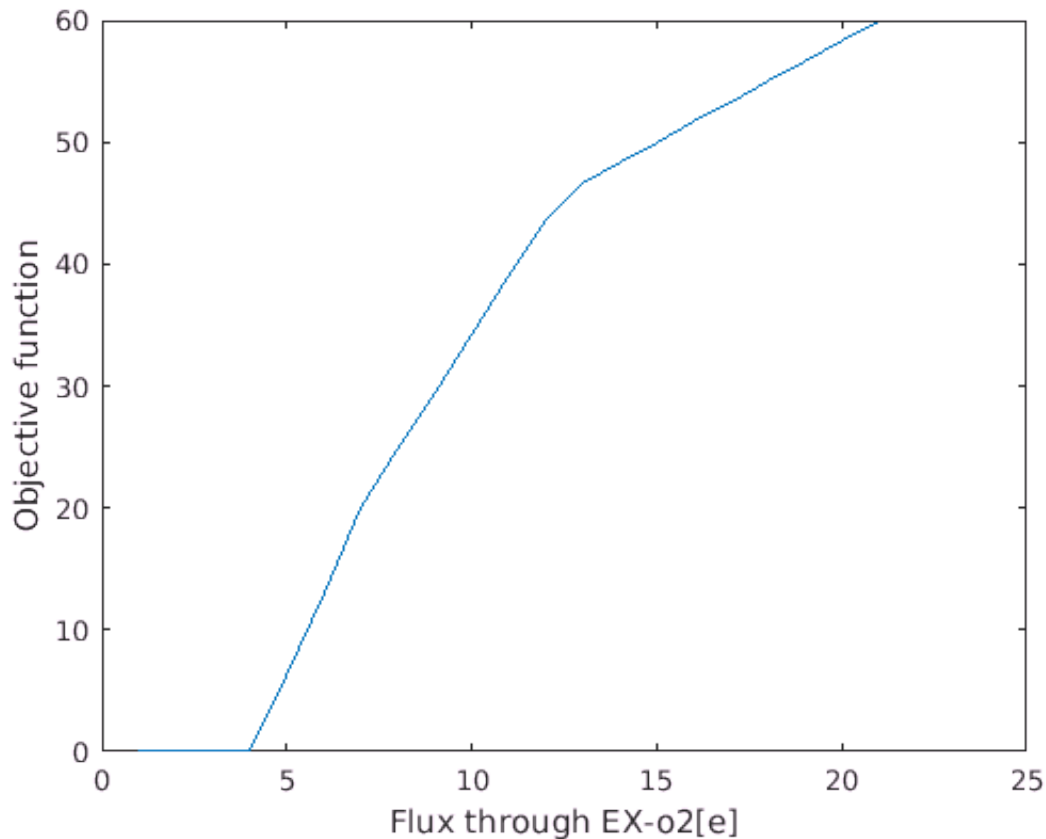


We can also investigate the robustness of the maximal ATP production when the available glucose amount is fixed, while different levels of oxygen are available.

```

modelrobustoxy = modelalter;
modelrobustoxy = changeRxnBounds(modelrobustoxy, 'EX_glc_D[e]', -20, 'b');
AtpRatesoxy = zeros(21, 1);
for i = 0:20
    modelrobustoxy = changeRxnBounds(modelrobustoxy, 'EX_o2[e]', -i, 'b');
    modelrobustoxy = changeObjective(modelrobustoxy, 'DM_atp_c_');
    FBArrobustoxy = optimizeCbModel(modelrobustoxy, 'max');
    AtpRatesoxy(i+1) = FBArrobustoxy.f;
end
plot(1:21, AtpRatesoxy)
xlabel('Flux through EX-o2[e]')
ylabel('Objective function')

```



- **Double robust analysis**

Performs robustness analysis for a pair of reactions of interest and an objective of interest. The double robust analysis is implemented with the function `doubleRobustnessAnalysis()`.

```
% [controlFlux1, controlFlux2, objFlux] = doubleRobustnessAnalysis(model,...
% controlRxn1, controlRxn2, nPoints, plotResFlag, objRxn, objType)
```

The inputs are a COBRA model, two reactions for the analysis and optional inputs:

```
%INPUTS
% model          COBRA model to analyse,
% controlRxn1    The first reaction for the analysis,
% controlRxn2    The second reaction for the analysis;
%
%OPTIONAL INPUTS
% nPoints        The number of flux values per dimension (Default = 20)
% plotResFlag     Indicates whether the result should be plotted (Default = true)
% objRxn         is objective to be used in the analysis (Default = whatever
%               is defined in model)
% objType        Direction of the objective (min or max)
%               (Default = 'max')
```

```
modeldrobustoxy = modelalter;
```

```

modeldrobustoxy = changeRxnBounds(modeldrobustoxy, 'EX_glc_D[e]', -20, 'l');
modeldrobustoxy = changeRxnBounds(modeldrobustoxy, 'EX_o2[e]', -17, 'l');
[controlFlux1, controlFlux2, objFlux] = doubleRobustnessAnalysis(modeldrobustoxy,...
    'EX_glc_D[e]', 'EX_o2[e]', 10, 1, 'DM_atp_c_', 'max')

```

Double robustness analysis in progress ...

1% []2% []3%

controlFlux1 =

```

-20.0000
-17.7778
-15.5556
-13.3333
-11.1111
-8.8889
-6.6667
-4.4444
-2.2222
0

```

controlFlux2 =

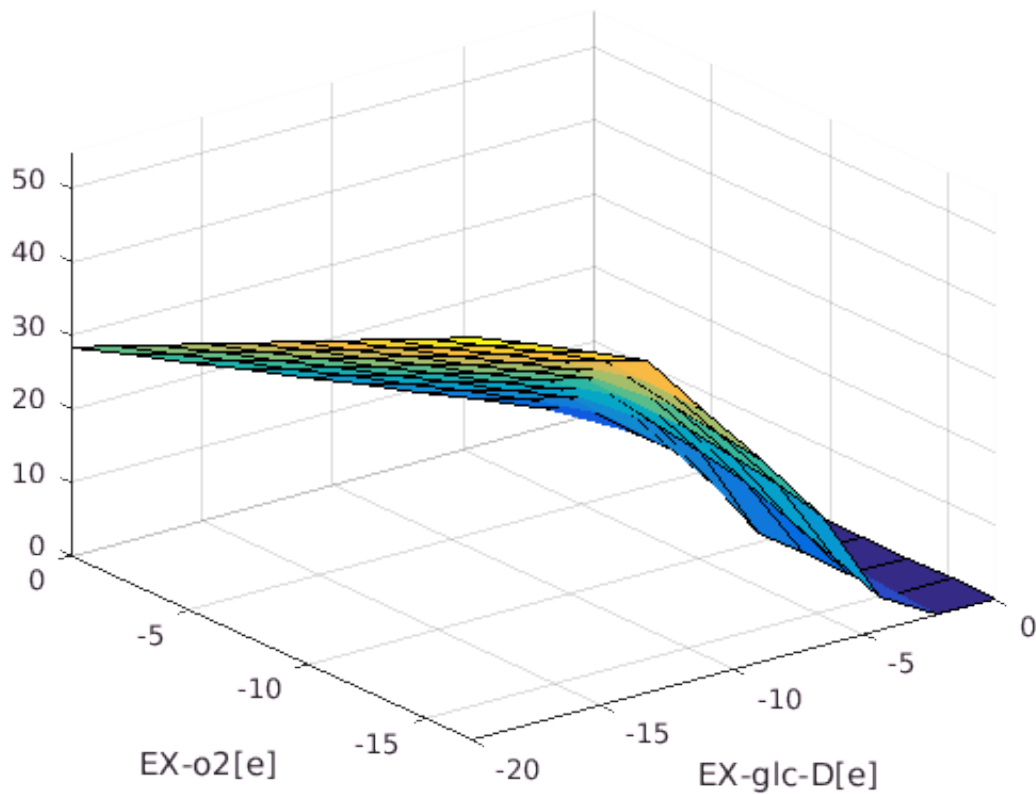
```

-17.0000
-15.1111
-13.2222
-11.3333
-9.4444
-7.5556
-5.6667
-3.7778
-1.8889
0

```

objFlux =

55.0000	51.8519	48.7037	45.2632	36.3158	27.3684	17.6923	4.6154
52.0370	48.8889	45.7407	42.5926	37.2515	28.3041	19.3567	7.0085
49.0741	45.9259	42.7778	39.6296	36.4815	29.2398	20.2924	9.4017
46.1111	42.9630	39.8148	36.6667	33.5185	30.1754	21.2281	11.7949
43.1481	40.0000	36.8519	33.7037	30.5556	27.4074	22.1637	13.2164
40.1852	37.0370	33.8889	30.7407	27.5926	24.4444	21.2963	14.1520
37.2222	34.0741	30.9259	27.7778	24.6296	21.4815	18.3333	15.0877
34.2593	31.1111	27.9630	24.8148	21.6667	18.5185	15.3704	12.2222
31.2963	28.1481	25.0000	21.8519	18.7037	15.5556	12.4074	9.2593
28.3333	25.1852	22.0370	18.8889	15.7407	12.5926	9.4444	6.2963



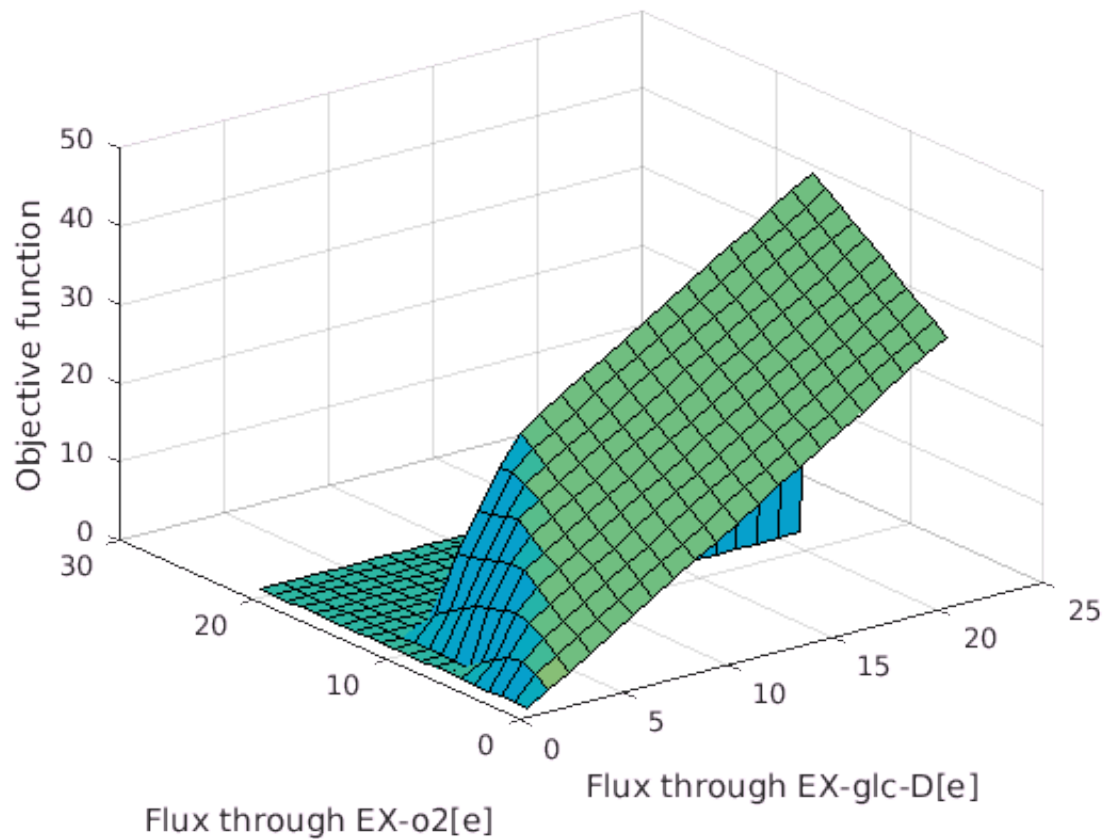
Phenotypic phase plane analysis (PhPP)

The PhPP is a method for describing in two or three dimensions, how the objective function would change if additional metabolites were given to the model [3].

Essentially PhPP performs a `doubleRobustnessAnalysis()`, with the difference that shadow prices are retained. The code is as follows-

```
modelphp = modelalter;
ATPphpRates = zeros(21);
for i = 0:10
    for j = 0:20
        modelphp = changeRxnBounds(modelphp, 'EX_glc_D[e]', -i, 'b');
        modelphp = changeRxnBounds(modelphp, 'EX_o2[e]', -j, 'b');
        modelphp = changeObjective(modelphp, 'DM_atp_c_');
        FBApph = optimizeCbModel(modelphp, 'max');
        ATPphpRates(i+1,j+1) = FBApph.f;
    end
end

surf1(ATPphpRates) % 3d plot
xlabel('Flux through EX-glc-D[e]')
ylabel('Flux through EX-o2[e]')
zlabel('Objective function')
```

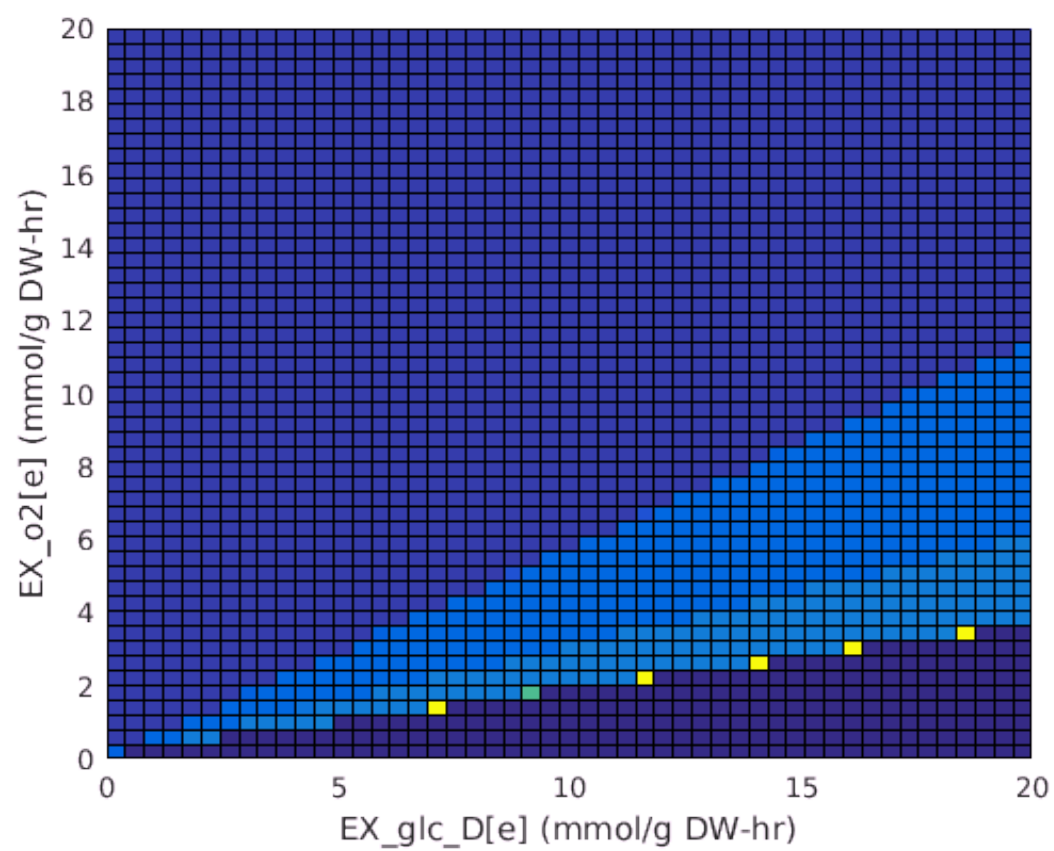
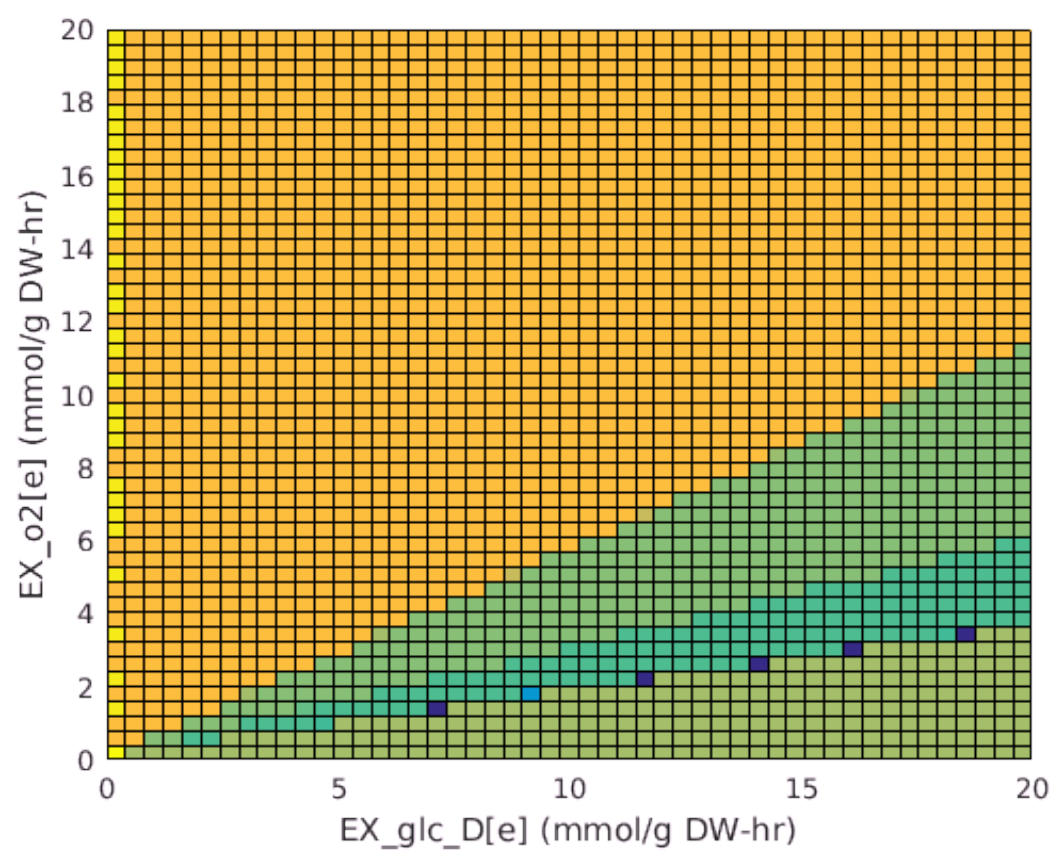



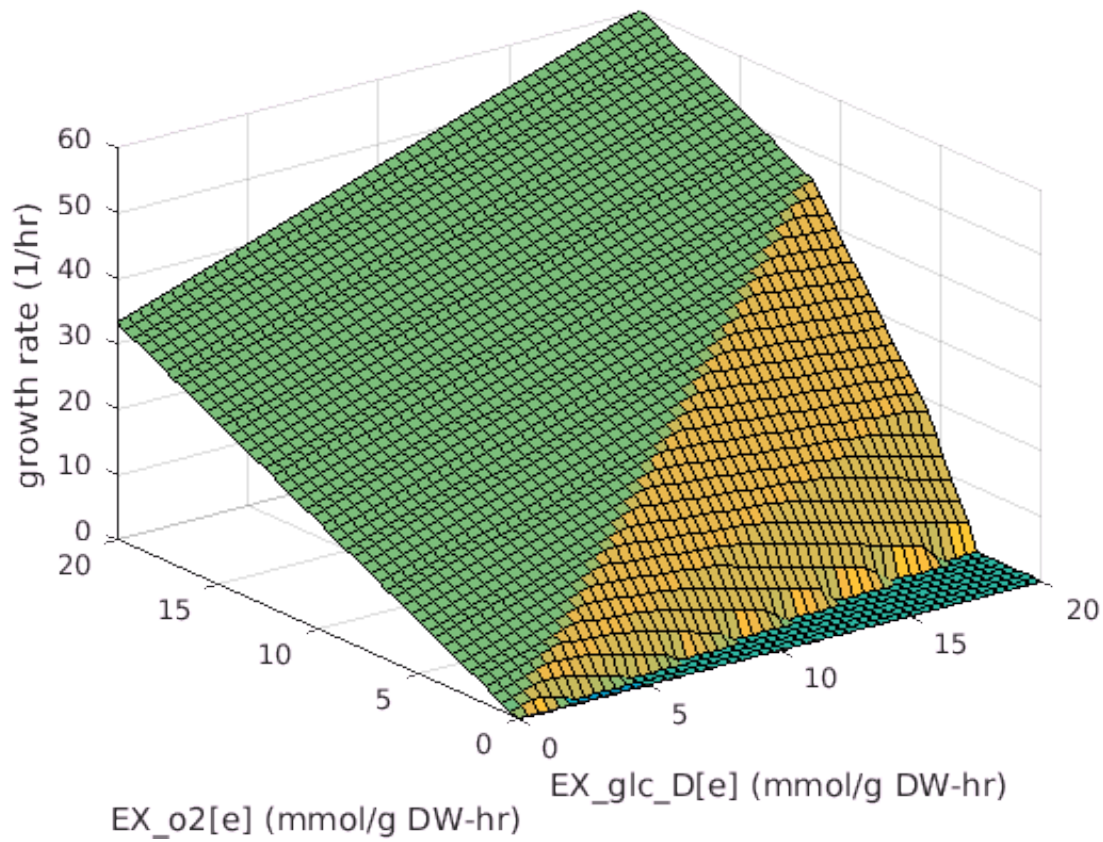
To generate a 2D plot: `pcolor(ATPphppRates)`

Alternatively, use the function `phenotypePhasePlane()`. This function also draws the line of optimality, as well as the shadow prices of the metabolites from the two control reactions. In this case, control reactions are 'EX_glc_D[e]' and 'EX_o2[e]'. The line of optimality signifies the state wherein, the objective function is optimal. In this case it is 'DM_atp_c_'.

```
modelphpp = changeObjective(modelphpp, 'DM_atp_c_');
[growthRates, shadowPrices1, shadowPrices2] = phenotypePhasePlane(modelphpp, ...
    'EX_glc_D[e]', 'EX_o2[e]');
```

generating PhPP





REFERENCES

- [1] Noronha A., et al. (2017). ReconMap: an interactive visualization of human metabolism. *Bioinformatics.*, 33 (4): 605-607.
- [2] Edwards, J.S. and and Palsson, B. Ø. (2000). Robustness analysis of the Escherichia coli metabolic network. *Biotechnology Progress*, 16(6):927-39.
- [3] Edwards, J.S., Ramakrishna, R. and and Palsson, B. Ø. (2002). Characterizing the metabolic phenotype: A phenotype phase plane analysis. *Biothechnology and Bioengineering*, 77:27-36.