# Solution to Exercise 8

#### Task 1

First, we read the *E. coli* core model into the Matlab workspace:

```
clear
model = readCbModel('e_coli_core.mat');
```

#### Task 2

Next, we want to determine a wild-type flux distribution w. To this end, we solve a FBA problem using the COBRA toolbox function optimizeCbModel.

```
% first, we can check the objective:
fprintf('Current objective: %s\n',model.rxnNames{model.c==1})

Current objective: Biomass Objective Function with GAM

% FBA
wt_sol = optimizeCbModel(model);
w = wt_sol.v;
```

### Task 3

Now, we come to the part, where we want to determine a flux for the mutant. Therfore, we first determine the significane thresholds for changes in flux for each reaction:

```
w_i^u=w_i+\delta \ |w_i|+\varepsilon // "significantly" higher w_i^l=w_i-\delta \ |w_i|-\varepsilon // "significantly" lower, where \delta=0.05 and \varepsilon=0.001.
```

```
delta = 0.05;
epsilon = 0.001;
% calculate thresholds
t_upper = w + delta * abs(w) + epsilon;
t_lower = w - delta * abs(w) - epsilon;
```

We need  $\varepsilon$  to account for possible numerical issues. For example, this means that even if the mutant flux exceeds the  $w_i + \delta |w_i|$  by a small values due to some numerical instability, it will not be recognized as significantly changed because we accounted for this error by adding a small value  $\varepsilon$  to the threshold.

#### Task 4

Now, we start constructing the mixed-integer linear problem (MILP) by forming the inequality constraints

$$v_i - y_i(v_{\text{max},i} - w_i^u) \le w_i^u$$
, and

```
v_i - y_i (v_{\min,i} - w_i^l) \ge w_i^l.
```

Here, we introduce binary variables  $y_i$  associated with each of the reactions. If  $y_i = 1$ , it means that reaction i is significantly changed either above or below the previously defined thresholds  $w_i^u$  and  $w_i^l$ .

We further have to transform the second inequality constraint into a lower-than-or-equal constraint by multiplying by -1:

$$-v_i + y_i \left( v_{\min,i} - w_i^l \right) \le -w_i^l.$$

The result inequality constraint matrix will thus have the dimensions  $2r \times 2r$ , where r is the number of reactions in the model. The matrix and associated right hand side of the of constaints will look like this:

$$\begin{bmatrix} 1 & 0 & \cdots & 0 & -v_{\max,1} + w_1^u & 0 & \cdots & 0 \\ 0 & 1 & \vdots & 0 & -v_{\max,2} + w_2^u & \vdots \\ \vdots & \ddots & 0 & \vdots & & \ddots & 0 \\ 0 & \cdots & 0 & 1 & 0 & \cdots & 0 & -v_{\max,r} + w_r^u \\ -1 & 0 & \cdots & 0 & v_{\min,1} - w_1^l & 0 & \cdots & 0 \\ 0 & -1 & \vdots & 0 & v_{\min,2} - w_2^l & & & & \\ \vdots & \ddots & 0 & \vdots & & \ddots & 0 \\ 0 & \cdots & 0 & -1 & 0 & & 0 & v_{\min,2} - w_2^l \end{bmatrix} \cdot \begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ v_r \\ y_1 \\ y_2 \\ \vdots \\ y_r \end{bmatrix} \leq \begin{bmatrix} w_1^u \\ w_2^u \\ \vdots \\ w_r^u \\ -w_1^l \\ -w_2^l \\ \vdots \\ -w_r^l \end{bmatrix}$$

It follows the implementation:

## Task 5

We proceed with the remaining parts of the MILP and the knock-out part.

To knock-out reaction NAD transhydrogenase we knockout genes b3962 and b1602 or b1603.

```
% knock-out genes of reaction 'NAD transhydrogenase'
% rule: 'b3962 or (b1602 and b1603)'
idx_b3962 = find(strcmp(model.genes, 'b3962'));
idx_b1602 = find(strcmp(model.genes, 'b1602'));
idx_b1603 = find(strcmp(model.genes, 'b1603'));
```

We also check which other reaction(s) may be affected by the knock-out.

```
rxns_related_to_b3962 = find(model.rxnGeneMat(:,idx_b3962)~=0);
```

```
rxns_related_to_b1602 = find(model.rxnGeneMat(:,idx_b1602)~=0);
rxns_related_to_b1603 = find(model.rxnGeneMat(:,idx_b1603)~=0);
model.grRules(unique([rxns_related_to_b3962; rxns_related_to_b1602; rxns_related_to_b1603]))
ans = 2×1 cell
'b1602 and b1603'
```

We find that these genes relate to one other reaction, checking the rules we find that this reaction will be blocked as well.

'b3962 or (b1602 and b1603)'

```
% using the COBRA function
ko_gene = {'b3962' 'b1602'};
ko rxns = struct2cell(findRxnsFromGenes(model,ko gene));
ko_rxns = find(ismember(model.rxns,cellfun(@(x) x(1),ko_rxns)));
% Equality constraints(all-zero for y-part)
Aeq = [
%
    model.S zeros(size(model.S));
];
beq = model.b;
% lower and upper bounds
%
        V
                    У
lb = [model.lb; zeros(r,1)];
ub = [model.ub; ones(r,1)];
% block reactions that are no longer active in the mutant
1b(ko rxns) = 0;
ub(ko_rxns) = 0;
% objective
       V
                    У
f = [zeros(r,1), ones(r,1)];
\% finally, we have to specify that the y varaibles are binary and not
% continuous (indices are "number of reactions until 2 times number of
% reactions)
intcon = r+1:2*r;
% solve the MILP to obtain the mutant flux distribution
options = optimoptions('intlinprog');
options.Display = 'off';
v = intlinprog(f,intcon,A,b,Aeq,beq,lb,ub,[],options);
% find the reactions that are significantly changed
changed_rxns = model.rxnNames(logical(v(intcon)));
disp('The following reaction changed significantly with KO of');...
disp(ko_gene);
```

The following reaction changed significantly with KO of

```
{'b3962'} {'b1602'}
```

```
arrayfun(@(i)fprintf('#%d\t\t%s\n',i,changed_rxns{i}),1:numel(changed_rxns))
```

- #1 Glucose-6-phosphate isomerase
- #2 Acetaldehyde dehydrogenase (acetylating)
- #3 Alcohol dehydrogenase (ethanol)
- #4 ATP maintenance requirement
- #5 Pyruvate transport in via proton symport
- #6 D lactate transport via proton symport
- #7 NAD(P) transhydrogenase
- #8 Formate exchange
- #9 Isocitrate lyase
- #10 D-lactate dehydrogenase
- #11 Malic enzyme (NADP)
- #12 NAD transhydrogenase

### Task 6

Finally, we calculate the Euclidean distance between the wild-type and the mutant flux vectors.

```
dist = sqrt(sum((w-v(1:r)).^2));
fprintf('The Euclidean distance between wild-type and mutant flux vectors is %.2f\n',dist)
```

The Euclidean distance between wild-type and mutant flux vectors is 9.70