

Optimization based on Integer Linear Programming

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Mathematical programming model

- In an *optimization problem*, the aim is to maximize (or minimize) a given quantity designated by the *objective* that depends on a finite number of variables.
- The variables might be independent or might be related between them through one or more *constraints*.
- A mathematical programming problem is an optimization problem such that the objective and the constraints are defined by mathematical functions and functional relations.
- A mathematical programming model describes a mathematical programming problem.

Mathematical programming model

For a given set of n variables $X = \{x_1, x_2, ..., x_n\}$, the standard way of defining a Mathematical Programming Model is:

Minimize (or Maximize)

Subject to:

$$g_i(X) \le k_i$$
 , $i = 1, 2, ..., m$ (=) (\geq)

where:

- -m is the number of constraints
- -f(X) and all $g_i(X)$ are functions of the variables
- $-k_i$ are real constants

(Mixed Integer) Linear Programming model

- A <u>Linear Programming</u> (LP) model is a mathematical programming model where all variables $X = \{x_1, x_2, ..., x_n\}$ are non-negative reals and f(X) and $g_i(X)$ are linear functions:
 - functions in the form $a_1x_1 + a_2x_2 + ... + a_nx_n$ where all a_i are real constants.
- An <u>Integer Linear Programming</u> (ILP) model is an LP model where all variables $X=\{x_1, x_2, ..., x_n\}$ are nonnegative integers.
- A <u>Mixed Integer Linear Programming</u> (MILP) model is an LP model where some variables $X=\{x_1, x_2, ..., x_n\}$ are nonnegative integers and others are non-negative reals.

Illustrative example

Consider a transportation company that has been requested to deliver the following items to a particular destination:

Item <i>i</i> :	1	2	3	4	5	6
Revenue (r_i) :	2.3	4.5	1.5	5.4	2.9	3.2
Size (s _i):	30	70	20	80	35	40

The company has 2 vans for item delivery:

- the first van has a capacity of 100
- the second van has a capacity of 60.

Since it is not possible to deliver all items with the 2 vans, the aim is to choose the items to be carried on each van to maximize the revenue.

Solving steps:

1st - define the ILP model of the optimization problem

2nd – solve the ILP model (using an available solver)

Illustrative example

Item <i>i</i> :	1	2	3	4	5	6
Revenue (r_i) :	2.3	4.5	1.5	5.4	2.9	3.2
Size (s_i) :	30	70	20	80	35	40

VARIABLES DEFINING THE PROBLEM:

- x1 Binary variable that, if is 1 in the solution, indicates that item 1 is delivered
- $_{ imes2}$ Binary variable that, if is 1 in the solution, indicates that item 2 is delivered

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- x6 Binary variable that, if is 1 in the solution, indicates that item 6 is delivered
- y1,1 Binary variable that, if is 1 in the solution, indicates that item 1 is carried by first van
- y1,2 Binary variable that, if is 1 in the solution, indicates that item 1 is carried by second van

•••

- y6,1 Binary variable that, if is 1 in the solution, indicates that item 6 is carried by first van
- y6,2 Binary variable that, if is 1 in the solution, indicates that item 6 is carried by second van

Illustrative example

Item <i>i</i> :	1	2	3	4	5	6
Revenue (r_i) :	2.3	4.5	1.5	5.4	2.9	3.2
Size (<i>s_i</i>):	30	70	20	80	35	40

The objective function is the total revenue

INTEGER LINEAR PROGRAMMING (ILP) MODEL (in LP format):

of the delivered items Maximize $+ 2.3 \times 1 + 4.5 \times 2 + 1.5 \times 3 + 5.4 \times 4 + 2.9 \times 5 + 3.2 \times 6$ Subject To + 30 y1,1 + 70 y2,1 + 20 y3,1 + 80 y4,1 + 35 y5,1 + 40 y6,1 <= 100 + 30 y1,2 + 70 y2,2 + 20 y3,2 + 80 y4,2 + 35 y5,2 + 40 y6,2 <= 60+ v1,1 + v1,2 - x1 = 0The total size of the items carried on each + v2,1 + v2,2 - x2 = 0van must be within the van capacity + y3,1 + y3,2 - x3 = 0+ y4,1 + y4,2 - x4 = 0+ y5, 1 + y5, 2 - x5 = 0 If an item is carried in one van, then, the + y6,1 + y6,2 - x6 = 0item is delivered Binary x1 x2 x3 x4 x5 x6 y1,1 y1,2 y2,1 y2,2 y3,1 y3,2 y4,1 y4,2 y5,1 y5,2 y6,1 y6,2 End

Illustrative example – using CPLEX (1)

Starting CPLEX:

```
Welcome to IBM(R) ILOG(R) CPLEX(R) Interactive Optimizer 12.6.1.0 with Simplex, Mixed Integer & Barrier Optimizers 5725-A06 5725-A29 5724-Y48 5724-Y49 5724-Y54 5724-Y55 5655-Y21 Copyright IBM Corp. 1988, 2014. All Rights Reserved.
```

Type 'help' for a list of available commands.

Type 'help' followed by a command name for more information on commands.

CPLEX>

Reading file 'exemplo.lp' on CPLEX:

```
CPLEX> read exemplo.lp
Problem 'exemplo.lp' read.
Read time = 0.01 sec. (0.00 ticks)
CPLEX>
```

Illustrative example – using CPLEX (2)

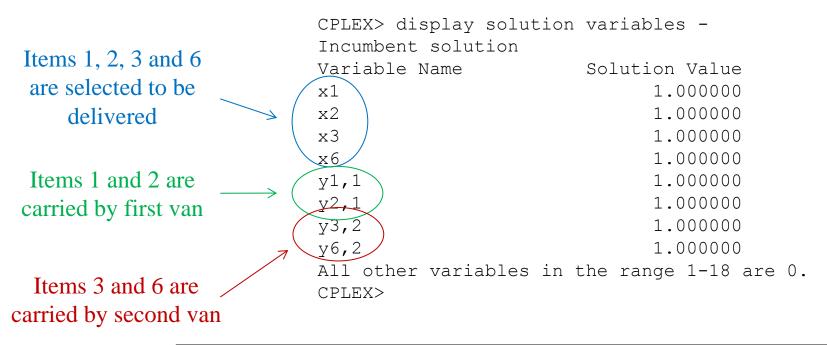
Solving the problem on CPLEX:

Deterministic time = 1.12 ticks (7.17 ticks/sec)

```
CPLEX> optimize
       Nodes
                                                   Cuts/
  Node Left
             Objective IInf Best Integer Best Bound
                                                                       Gap
                                                            ItCnt
                                      1.5000
                                                   19.8000
     0+
     0+
                                     11.2000
                                                   19.8000
                                                                    76.79%
     0
           0
                  11.8286 1
                                    11.2000
                                                   11.8286
                                                                 4 5.61%
                                                                      2.86%
                                     11.5000
                                                   11.8286
     0+
                                     11.5000
                  cutoff
                                                                      0.00%
Elapsed time = 0.14 sec. (1.12 ticks, tree = 0.00 MB, solutions = 3)
Root node processing (before b&c):
 Real time
                      = 0.14 sec. (1.12 ticks)
Parallel b&c, 4 threads:
                     = 0.00 sec. (0.00 ticks)
 Real time
 Sync time (average) = 0.00 \text{ sec.}
 Wait time (average) = 0.00 sec.
                                                             Optimal solution value
Total (root+branch&cut) = 0.14 sec. (1.12 ticks)
Solution pool: 4 solutions saved.
MIP - Integer optimal solution: Objective \(\pm\) 1.1500000000e+001
Solution time = 0.16 sec. Iterations = 4 Nodes = 0
                                                                                9
```

Illustrative example – using CPLEX (3)

Displaying the values of the optimal solution:



Item i:	1	2	3	4	5	6
Revenue (r_i) :	2.3	4.5	1.5	5.4	2.9	3.2
Size (s _i):	30	70	20	80	35	40

Illustrative example – mathematical notation

Parameters:

n – number of items r_i – revenue of delivering item i, with i = 1, ..., n

 s_i – size of item i, with i = 1,...,n

v – number of vans c_j – capacity of van j, with j = 1, ..., v

Variables:

 x_i – binary variable that is 1 if item i is delivered, i = 1,...,n

 y_{ij} – binary variable that is 1 if item i is carried on van j, i = 1,...,n and j = 1,...,v

ILP model: Maximize
$$\sum_{i=1}^{n} r_i x_i$$

Subject to:

$$\sum_{i=1}^{n} s_i y_{ij} \le c_j$$
 , $j = 1 \dots v$
 $\sum_{j=1}^{v} y_{ij} = x_i$, $i = 1 \dots n$
 $x_i \in \{0,1\}$, $i = 1 \dots n$
 $y_{ij} \in \{0,1\}$, $i = 1 \dots n$, $j = 1, \dots v$

Illustrative example – generating LP file with MATLAB

```
c = [100 60];
                                                     n= length(r);
                                                     v= length(c);
                                                     fid = fopen('exemplo.lp','wt');
                                                     fprintf(fid, 'Maximize\n');
                   Maximize \sum_{i=1}^{n} r_i x_i for i=1:n fprintf(fid,' + %f x%d',r(i),i);
                                                     fprintf(fid, '\nSubject To\n');
                                                     for j=1:v
                                                          for i=1:n
        \sum_{i=1}^{n} s_i y_{ij} \le c_j \quad , j = 1 \dots v - 
                                                                fprintf(fid,' + %f y%d,%d',s(i),i,j);
                                                           fprintf(fid,' <= %f\n',c(j));
                                                    for i=1:n
         \sum_{j=1}^{v} y_{ij} = x_i \quad \text{,} i = 1 \dots n \xrightarrow{\text{for } j=1:v \\ \text{fprintf(fid,' + y%d,%d',i,j);}}} \text{end}
                                                           fprintf(fid,' - x%d = 0 \n', i);
                                                     fprintf(fid, 'Binary\n');
                                                     for i=1:n
x_i \in \{0,1\} \text{, } i=1\dots n \text{ fprintf(fid,' x%d\n',i); }  y_{ij} \in \{0,1\} \text{, } i=1\dots n \text{ , } j=1,\dots v \text{ fprintf(fid,' x%d\n',i); } 
                                                                fprintf(fid, ' y%d, %d\n', i, j);
                                                                                                         12
                                                     fprintf(fid, 'End\n');
```

fclose(fid);

 $r = [2.3 \ 4.5 \ 1.5 \ 5.4 \ 2.9 \ 3.2];$

 $s = [30 \ 70 \ 20 \ 80 \ 35 \ 40];$

Illustrative example - using Gurobi on Internet (1)

 Prepare an ASCII file with the problem defined in LP format and compress it with Zip:

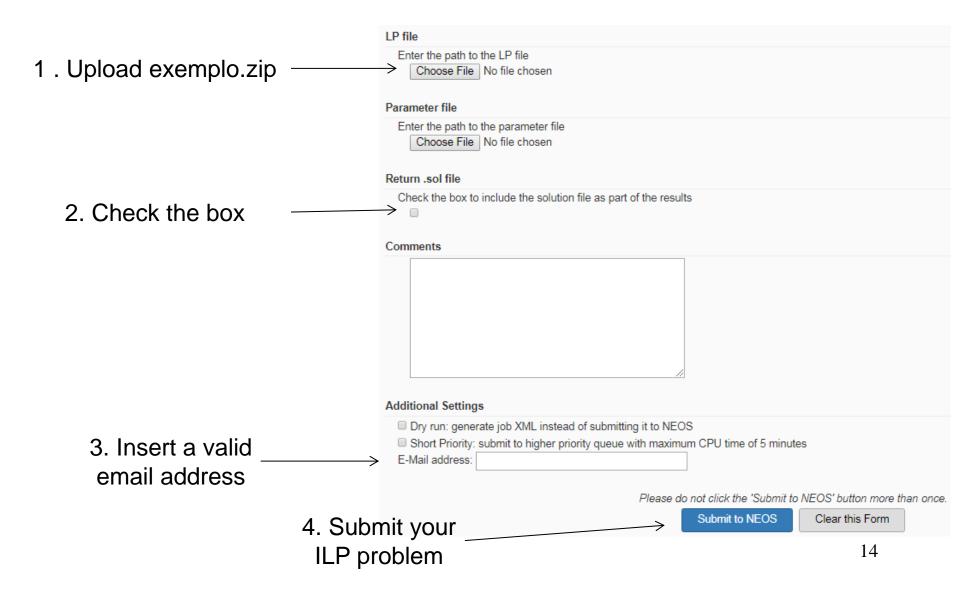
for example: exemplo.zip

- Go to https://neos-server.org/neos/solvers/index.html
- Select Mixed Integer Linear Programming tools
- Select Gurobi [LP Input]

Mixed Integer Linear Programming

- · Cbc [AMPL Input][GAMS Input][MPS Input]
- CPLEX [AMPL Input][GAMS Input][LP Input][MPS Input]
- · feaspump [AMPL Input][CPLEX Input][MPS Input]
- FICO-Xpress [AMPL Input][GAMS Input][MOSEL Input][MPS Input]
- · Gurobi [AMPL Input][GAMS Input][LP Input][MPS Input]
- MINTO [AMPL Input]
- MOSEK [AMPL Input][GAMS Input][LP Input][MPS Input]
- proxy [CPLEX Input][MPS Input]
- qsopt_ex [AMPL Input][LP Input][MPS Input]
- scip [AMPL Input][CPLEX Input][GAMS Input][MPS Input][OSIL Input][ZIMPL Input]
- SYMPHONY [MPS Input]

Illustrative example - using Gurobi on Internet (2)



Illustrative example - using Gurobi on Internet (3)

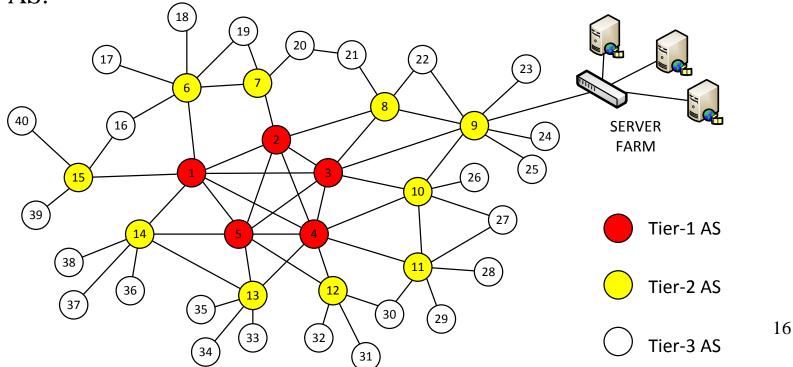
 After the problem is solved, the solution is displayed (and also sent to the email address):

```
Optimal solution found (tolerance 1.00e-04)
Best objective 1.150000000000e+01, best bound 1.15000000000e+01, gap 0.0000%
Optimal objective: 11.5
****** Begin .sol file *******
# Objective value = 11.5
x1 1
x2 1
x3 1
x4 0
x5 0
x6 1
y1,11
y2,11
y3,10
y4,10
y5,10
y6,10
y1,20
y2,20
y3,2 1
y4,20
y5,20
v6,2 1
```

****** End .sol file *******

Solving the server farm location problem with ILP

- We have a set of Autonomous Systems (ASs) and we aim to select a subset of ASs to connect one server farm on each selected AS.
- Only Tier-2 of Tier-3 ASs provide the Internet access service.
- The solution must guarantee that there is a path between each Tier-2 and Tier-3 AS and at least one server farm with no more than one intermediate AS.



Server farm location problem: Notation and Variables

NOTATION:

n – number of Tier-2 and Tier-3 ASs where server farms can be connected to;

 c_i – OPEX cost of connecting a server farm to AS i, with $1 \le i \le n$;

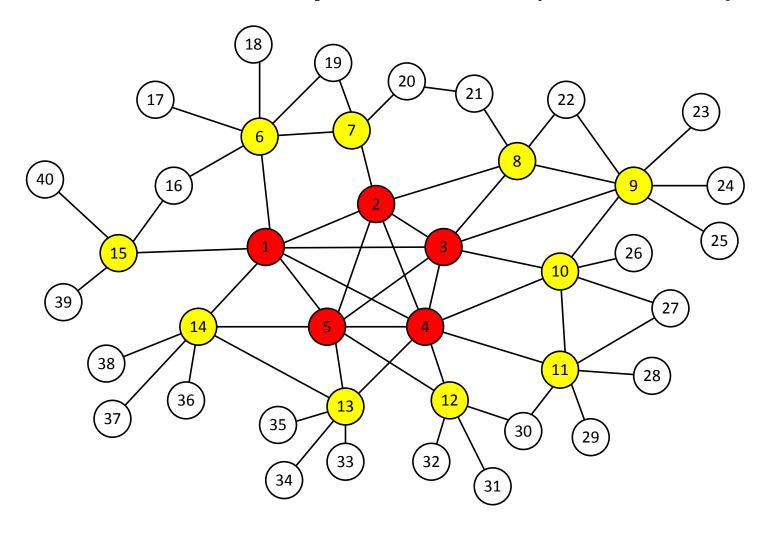
I(j) – set of Tier-2 and Tier-3 ASs such that there is a shortest path between AS j and each AS $i \in I(j)$ with at most one intermediate AS.

VARIABLES:

 x_i – binary variable, with $1 \le i \le n$, that when is equal to 1 means that AS i must be connected to one server farm;

 y_{ji} – binary variable, with $1 \le j \le n$ and $i \in I(j)$, that when is equal to 1 means that AS j is associated with AS i.

Server farm location problem: examples of sets I(j)



Set I(j) for j = 6 is: $\{6,7,14,15,16,17,18,19,20\}$ for j = 16 is: $\{6,7,15,16,17,18,19,39,40\}$

Server farm location problem: ILP Model

$$Minimize \sum_{i=1}^{n} c_i x_i$$
 (1)

Subject to:

$$\sum_{i \in I(j)} y_{ji} = 1 \qquad , j = 1 \dots n$$
 (2)

$$y_{ii} \le x_i$$
 , $j = 1 ... n, i \in I(j)$ (3)

$$x_i \in \{0,1\}$$
 , $i = 1 \dots n$ (4)

$$y_{ji} \in \{0,1\}$$
 , $j = 1 ... n, i \in I(j)$ (5)

- The objective (1) is the minimization of the OPEX costs of the selected server farms.
- Constraints (2) guarantee that each AS j is associated with one AS $i \in I(j)$ while constraints (3) guarantee that an associated AS $i \in I(j)$ must have one server farm connected. So, constraints (2–3) guarantee that each AS j has always one server farm whose shortest path has at most one intermediate AS.
- Constraints (4–5) define all variables as binary variables.