





CMPL <Coin | Coliop > Mathematical Programming Language

Mike Steglich

Technical University of Applied Sciences Wildau

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1 Introduction

- CMPL (<Coliop | Coin> Mathematical Programming Language) is a mathematical programming language and a system for mathematical programming and optimisation of linear optimisation problems. CMPL executes CBC (default), CPLEX, GLPK, Gurobi and SCIP directly to solve the generated model instance.
- o The CMPL distribution contains **Coliop** which is an IDE (Integrated Development Environment) for CMPL.
- pyCMPL is the CMPL application programming interface (API) for Python and an interactive shell and jCMPL is CMPL's Java API.
- **CMPLServer** is a XML-RPC-based web service for distributed and grid optimisation that can be used with CMPL, pyCMPL and jCMPL.
- The CMPL distribution is an Open Source Project and available for most of the relevant operating systems (Windows, OS X, Linux and Raspbian).
- CMPL, Coliop, pyCMPL, jCMPL and CMPLServer are COIN-OR projects initiated by the Technical University of Applied Sciences Wildau and the Institute for Operations Research and Business Management at the Martin Luther University Halle-Wittenberg.
- CMPL and pyCMPL are also available in SolverStudio, which is an add-in for Excel 2007 and later on Windows created by Andrew Mason (University Auckland, NZ) that allows a user to build and solve optimisation models in Excel optimisation modelling languages.

2 Selected language elements and features

2.1 Basic language elements

$$\sum_{i \in S} \sum_{j \in D} c_{ij} \cdot x_{ij} \to \min!$$
s.t.

$$\sum_{i \in D} x_{ij} = s_i \qquad ; i \in S$$

$$\sum_{i \in S} x_{ij} = d_j \quad ; j \in D$$

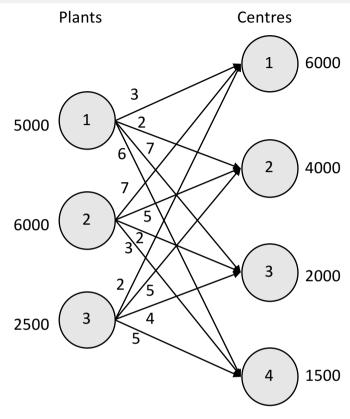
$$x_{ij} \ge 0$$
 ; $i \in S, j \in D$

Parameters:

- S set of the sources
- D set of the destinations
- s_i supply in units of source i
- d_i demand in units of destination j
- c_{ii} transportation costs per unit if a unit is shipped from source i to the destination j



 x_{ii} units that are to be transported from source i to the destination j



[Anderson et al. (2014), p. 281.]

[cf. Hillier and Liebermann (2010), p. 305ff, Williams (2013), p. 82ff, Vanderbei (2014), p. 225ff.]

2.1 Basic language elements

$$\sum_{i \in S} \sum_{j \in D} c_{ij} \cdot x_{ij} \to \min!$$

$$\sum_{j \in D} x_{ij} = s_i \quad ; i \in S$$

$$\sum_{i \in S} x_{ij} = d_j \quad ; j \in D$$

$$\sum_{i \in S} x_{ij} = d_j \quad ; j \in L$$

$$x_{ij} \ge 0$$
 ; $i \in S, j \in D$

parameters: S:=1..3;

$$D:=1..4;$$

$$c[S,D] := ((3,2,7,6), (7,5,2,3), (2,5,4,5));$$

$$s[S] := (5000,6000,2500);$$

$$d[D] := (6000, 4000, 2000, 1500);$$

variables:

objectives:

constraints:

2.1 Basic language elements

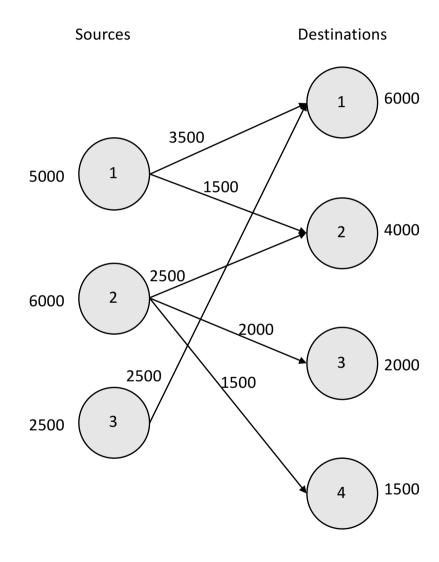
> **cmpl** transportation.cmpl

Problem	transportation.cmpl
Nr. of variables	12
Nr. of constraints	7
Objective name	costs
Solver name	CBC
Display variables	nonzero variables (all)
Display constraints	nonzero constraints (all)

Objective status optimal
Objective value 39500 (min!)

Vai	iables				
Nar	е Туре	Activity	Lower bound	Upper bound	Marginal
x[]	.11 C	3500	0	Infinity	
	,2] C		0	Infinity	0
x[2	,2] C	2500	0	Infinity	0
x[2	,3] C	2000	0	Infinity	0
x[2	,4] C	1500	0	Infinity	0
x[3	,1] C	2500	0	Infinity	0

Constraints					
Name	Туре	Activity	Lower bound	Upper bound	Marginal
supply[1]	Е	5000	5000	5000	1
supply[2]	E	6000	6000	6000	4
supply[3]	E	2500	2500	2500	-
demand[1]	E	6000	6000	6000	2
demand[2]	E	4000	4000	4000	1
demand[3]	E	2000	2000	2000	-2
demand[4]	E	1500	1500	1500	-1



[Anderson et al. (2014), p. 284.]

2.2 CMPL header and CmplData

CMPL header

A CMPL header is intended to define CMPL options, solver options and display options for the specific CMPL model. An additional intention of the CMPL header is to specify external data files which are to be connected to the CMPL model. The elements of the CMPL header are not part of the CMPL model and are processed before the CMPL model is interpreted.

CmplData

 A CmplData file is a plain text file that contains the definition of parameters and sets with their values in a specific syntax. The parameters and sets can be read into a CMPL model by using the CMPL header argument %data.

%arg -solver glpk %arg -cmplUrl ↓ http://194.95.44.187:8008	GLPK is to be executed on a CmplServer located at 194.95.44.187:8008
%opt cbc ratio 0.1	If CBC is chosen then a mipGap of 0.1 is used
%display var x*	Only variables starting with \boldsymbol{x} are to be displayed in the solution report.

2.2 CMPL header and CmplData

transportation.cdat

transportation.cmpl

2.3 List of the language elements

- Objects
 - Parameters
 - Variables
 - Indices and sets
 - Line names
- CMPL header
- Parameter Expressions
 - Array functions
 - Set operations and functions
 - Mathematical functions
 - Type casts
 - String operations
- Input and output operations
 - Error and user messages
 - cmplData files
 - Readcsv and readstdin
 - Include

- Control structures
 - For loop
 - If-then clause
 - Switch clause
 - While loop
 - Set and sum control structures as expression
- Matrix-Vector notations
- Automatic model reformulations
 - Matrix reductions
 - Equivalent transformations of products of variables

3 Application programming interfaces (pyCmpl and jCmpl)

- pyCMPL is the CMPL API for Python and an interactive shell and jCMPL is CMPL's Java API.
 - The main idea of the APIs are:
 - to define sets and parameters within the user application,
 - to start and control the solving process and
 - to read the solution(s) into the application if the problem is feasible.
 - All variables, objective functions and constraints are defined in CMPL.
 - These functionalities can be used with a local CMPL installation or a CMPLServer.

Application programming interfaces (pyCMPL and jCMPL)

Cost per machine and location

		Locations			
		1	2	3	4
səu	1	13	16	12	11
Machines	2	15	-	13	20
Ma	3	5	7	10	6

(Hillier/Liebermann 2010, p. 334f.)

Parameters:

N₁ Set of the machines

 N_2 Set of the locations

assignment.cmpl

```
%data : machines set, locations set, A set[2], c[A]
variables:
   x[A]: real[0..];
objectives:
   sum\{ [i,j] in A : C[i,j] *x[i,j] \} \rightarrow min ;
constraints:
   { i in machines: sum\{ j in (A *> [i,*]) : x[i,j] \} = 1; }
    { j in locations: sum{ i in (A *> [*,j]) : x[i,j] } <= 1;}
```

3 Application programming interfaces (pyCMPL and jCMPL)

Creating Cmpl object and defining sets and parameters (pyCMPL)

```
m = Cmpl("assignment.cmpl")
locations = CmplSet("locations")
locations.setValues(1,4)

machines = CmplSet("machines")
machines.setValues(1,3)

combinations = CmplSet("A", 2)
combinations.setValues([[1,1],[1,2],[1,3], ...]))

c = CmplParameter("c",combinations)
c.setValues([13,16,12,11,15,13,20,5,7,10,6])

m.setSets(machines,locations,combinations)
m.setParameters(c)
```

Solving and manipulating models (pyCMPL)

```
m.setOption("%arg -solver gurobi")
m.solve()
```

3 Application programming interfaces (pyCMPL and jCMPL)

Reading optimal solution

```
print m.solution.value
print m.solution.status

for v in m.solution.variables:
    print v.name, v.type, v.activity, v.lowerBound, v.upperBound

for c in m.solution.constraints:
    print c.name, c.type, c.activity, c.lowerBound, c.upperBound
```

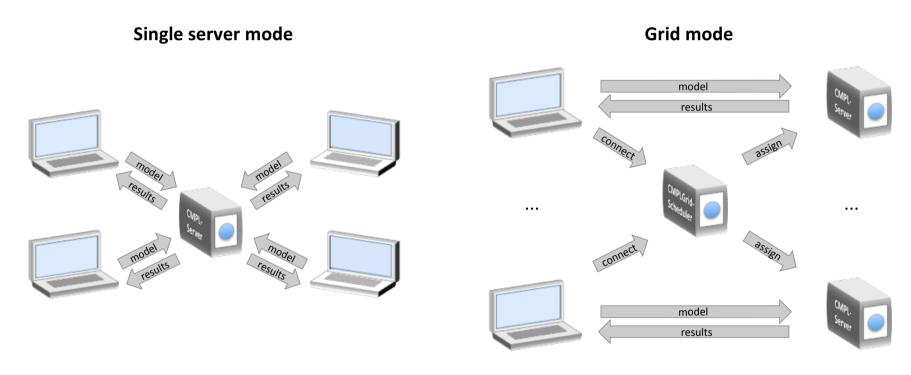
3 Application programming interfaces (pyCMPL and jCMPL)

assignment.py

```
#!/usr/bin/python
from pyCmpl import *
try:
    m = Cmpl("assignment.cmpl")
    locations = CmplSet("locations")
     locations.setValues(1,4)
    machines = CmplSet("machines")
    machines.setValues(1,3)
     combinations = CmplSet("A", 2)
     combinations.setValues([[1,1],[1,2],[1,3],[1,4],[2,1],[2,3],[2,4],[3,1],[3,2],[3,3],[3,4]])
     c = CmplParameter("c", combinations)
     c.setValues([13,16,12,11,15,13,20,5,7,10,6])
    m.setSets(machines, locations, combinations)
     m.setParameters(c)
    m.setOption("%arg -solver gurobi")
    m.solve()
     print m.solution.value
     print m.solution.status
     for v in m.solution.variables:
          print v.name, v.type, v.activity, v.lowerBound, v.upperBound
     for c in m.solution.constraints:
          print c.name, c.type, c.activity, c.lowerBound, c.upperBound
except CmplException, e:
     print e.msg
```

4 CMPLServer

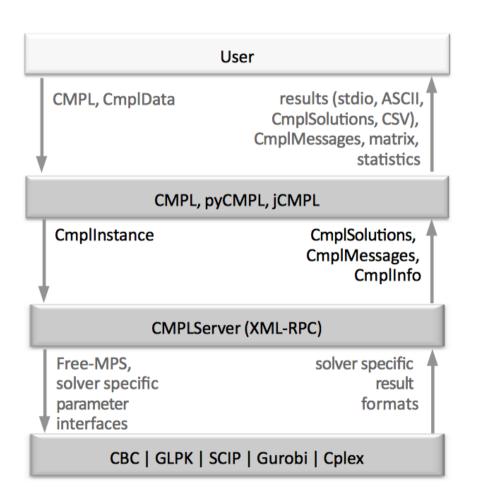
• The **CMPLServer** is an **XML-RPC-based web service** (cf. St. Laurent et al. 2001, p. 1.) for distributed and grid optimisation with clients for CMPL, pyCMPL and jCMPL that can be used in two modes:



- Both modes can be understood as distributed systems "in which hardware and software components located at networks computers communicate and coordinate their actions only by passing messages". (Coulouris et al, 2012, p. 17)
- Distributed optimisation is in this meaning interpretable as a distributed system that can be used for solving optimisation problems. (cf. Kshemkalyani & Singhal, 2008, p. 1; Fourer et.al., 2010)

4.1 Introduction

The communication between the clients and the server(s) works through XML-RPC and four CMPL-specific XML formats for the communication between clients and servers.



CmplInstance

XML file that contains all relevant information about a CMPL model sent to a CMPLServer

CmplMessages

XML file that contains the status and messages of a CMPL model

CmplSolutions

CMPL's XML based solution file format

CmplInfo

XML file that contains (if requested) several statistics and the generated matrix of the CMPL model

(XSD schemes: coliop.org/schemes)

4.2 Single server mode

```
cmplServer -start # CMPLServer IP 10.0.1.52 port 8008
```

CMPL synchronously

```
cmpl test.cmpl -cmplUrl http://10.0.1.52:8008
```

pyCMPL asynchronously

```
m = Cmpl("test.cmpl")
m.connect("http://10.0.1.52:8008")
m.send()
m.knock()
m.retrieve()
...
```

jCMPL synchronously

```
m = Cmpl("test.cmpl");
m.connect("http://10.0.1.52:8008");
m.solve();
...
```

```
cmplServer -stop
```

5 User interfaces (Coliop and SolverStudio)

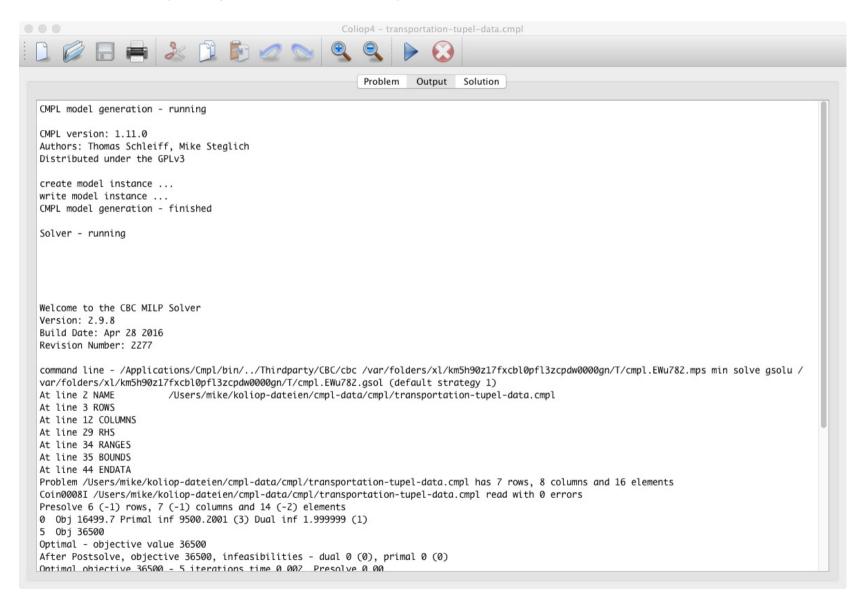
Coliop

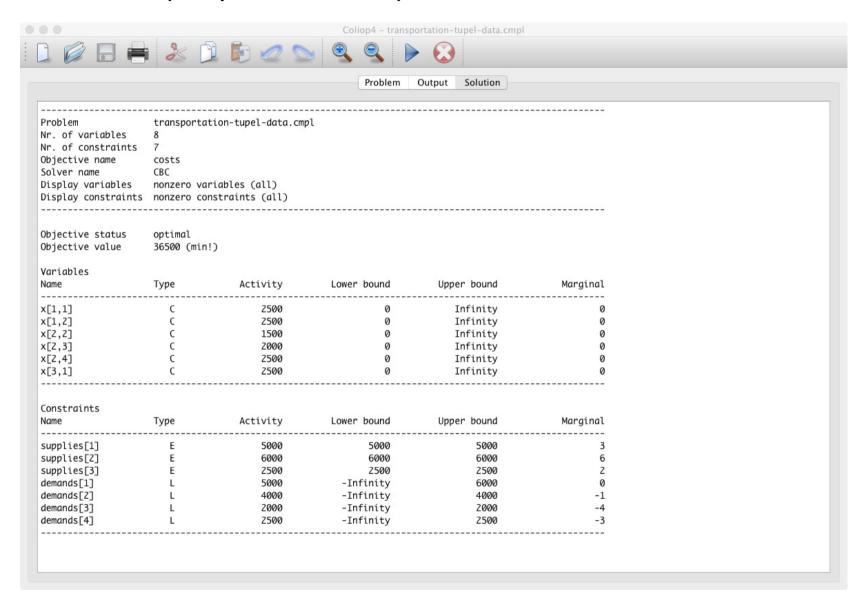
Coliop is an IDE (Integrated Development Environment) for CMPL intended to solve linear programming (LP) problems and mixed integer programming (MIP) problems. Coliop is an open source project licensed under GPL. It is written in C++ and is as an integral part of the CMPL distribution available for most of the relevant operating systems (OS X, Linux and Windows).

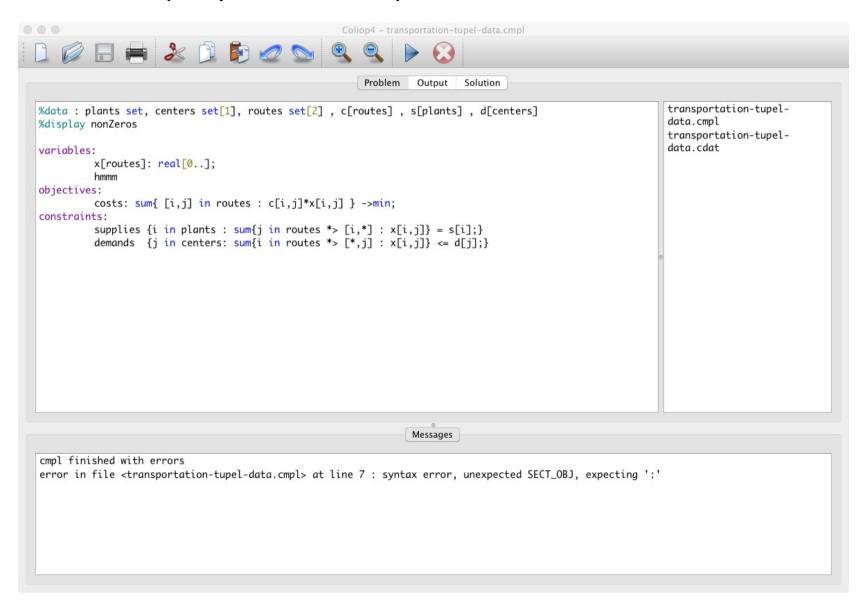
SolverStudio

- SolverStudio is an add-in for Excel on Windows that allows you to build and solve optimisation models in Excel using any of the following optimisation modelling languages:
 - PuLP, COOPR/Pyomo, AMPL, AMPL on NEOS server, GMPL (GNU MathProg Language), GAMS using the Gurobi Python modelling interface, SimPy, Julia and ...
 - CMPL and pyCMPL.
- SolverStudio allows you to create and edit your optimisation model without leaving Excel, and to save your model inside your workbook. You can also easily link data on your spreadsheet to sets, parameters, constants and variables used in the model. SolverStudio can run the model to solve the problem and then put the answer back onto the spreadsheet.

```
Coliop4 - transportation-tupel-data.cmpl
Problem Output Solution
                                                                                                             transportation-tupel-data.cmpl
%data : plants set, centers set[1], routes set[2] , c[routes] , s[plants] , d[centers]
                                                                                                             transportation-tupel-data.cdat
%display nonZeros
variables:
          x[routes]: real[0..];
objectives:
          costs: sum{[i,j] in routes : c[i,j]*x[i,j] } ->min;
 constraints:
          supplies {i in plants : sum{j in routes *> [i,*] : x[i,j]} = s[i];}
          demands {j in centers: sum\{i \text{ in routes *> [*,j] : }x[i,j]\} <= d[j];}
```







References

- Anderson, R. David, D.J. Sweeney, Th.A. Williams and M. Wisniewski (2014): An Introduction to Management Science - Quantitative Approaches to Decision Making, 2nded., Cengage Learning EMEA, Andover.
- Coulouris, G.F.; J. Dollimore, T. Kindberg, G. Blai (2012): Distributed Systems: Concepts and Design, 5th ed., Addison-Wesley.
- o Fourer, R, J. Ma, R. K. Martin (2010): Optimization Services: A Framework for Distributed Optimization. Operations Research 58(6), 1624-1636.
- Hillier, F.S. und G.J. Lieberman (2010): Introduction to Operations Research, 9th ed., McGraw-Hill, New York et al.
- Kshemkalyani, A.D., M. Singhal, M. (2008): Distributed Computing Principles, Algorithms, and Systems,
 Kindle ed., Cambridge University Press.
- Rogge, R. and M. Steglich (2007): Betriebswirtschaftliche Entscheidungsmodelle zur Verfahrenswahl sowie Auflagen- und Lagerpolitiken, in: Diskussionsbeiträge zu Wirtschaftsinformatik und Operations Research 10, Martin-Luther-Universität Halle-Wittenberg.
- St. Laurent, S., J. Johnston, E. Dumbill (2001): Programming Web Services with XML-RPC, 1st ed., O'Reilly.
- Vanderbei, R.J. (2014): Linear Programming: Foundations and Extensions, 4th ed., Springer, New York et al.
- Williams, H.P. (2013): Model Building in Mathematical Programming, 5th ed., Wiley, Chichester.

Thank you for your attention.

If you have any questions, please feel free to ask.