

# FLOPC++

*An algebraic modeling language embedded in C++*

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# FLOPC++



## Formulation of Linear Optimization Problems in C++

- Algebraic modelling language.
- Implemented as a C++ class library.
- Declarative optimization modelling (similar to GAMS, AMPL and AIMMS), within a C++ program.
- LP's and MIP's
- Open source (Common Public License version 1.0)
- <https://projects.coin-or.org/FlopC++>
- New users are appreciated.

# Overview of the talk



1. Motivation
2. How to use FLOPC++ (tips and tricks)
3. Implementation
4. Future plans (Stochastic Programming extensions)
5. Conclusion

# Traditional AMLs



## Strengths ...

- Syntax close to the notation used by most modellers
- Instance generation is automated
- Easy to modify models
- Representation is readable by both humans and computers

## Weaknesses ...

- Hard to integrate with other software components
- Limited procedural support for algorithm development
- Limited model/program structuring facilities
- Limited flexibility and extendibility

# Embedding an opt. model ...



in a C/C++ application.

## 1. Using the callable library of the solver.

```
...
// code to generate a problem instance
// in the format of the solver

lp.copylpdata(n, m, minimize, c, b, sense,
              Cst, Clg, Rnr, Elm, l, u);

solstat = lp.primopt();

if (solstat==CPX_OPTIMAL) lp.getX(solution);
...
```

# Embedding an opt. model ...



in a C/C++ application.

- **2. Using an algebraic modelling language.**
  1. Write code to generate a data file.
  2. Write code to spawn the algebraic modelling language interpreter.
  3. Write code to parse the results from a file.

# Declarative + procedural



- column generation
- decomposition
- heuristics
- etc.

We must combine declarative and procedural modelling.

Two approaches:

1. Add procedural constructs to an algebraic modelling language.
2. Add declarative modelling constructs to a general purpose programming language.

# Other modelling libraries



**'MP' (Søren S. Nielsen) C++**

**Planner (ILOG) C++**

**Concert Technology (ILOG) C++**

**LP Toolkit (Euro-decision) C++**

**XBSL (Dash) C**

**AMMO (Drayton Analytics) COM**

**Component Libraries (ILOG) COM, based on OPL**

**OptiMax2000 (Maximal Software) COM, based on MPL**

**EMOSL (Dash) C, based on XPRESS-MP**



# Comparison I



## ILOG Concert Technology

```
for (IloInt s=0; s<numS; s++)  
    model.add( IloSum(x[s]) <= capacity[s] );
```

## FLOPC++

```
supply(S) = sum(D, x(S,D)) <= capacity(S);
```

# Comparison II



## ILOG Concert Technology

```
for (IloInt d=0; d<numD; d++) {  
    IloExpr total(env);  
    for (IloInt s=0; s<numS; s++) {  
        total += x[s][d];  
    }  
    model.add( total >= dem[d] );  
    total.end();  
}
```

## FLOPC++

```
demand(D) = sum(S, x(S,D)) >= dem(D);
```

# Non-Commercial



From the web pages of Karlsruhe Universität:

**WiOR**

Universität Karlsruhe (TH) - Institut für  
Wirtschaftstheorie und Operations Research

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*Modeling Languages - Non-Commercial Software*

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Unfortunately we currently have no packages in the list.

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# An example



```
main() {
  MP_set S(numS),D(numD);
  MP_subset<2> Link(S,D);
  MP_data SUPPLY(S), DEMAND(D), d(Link), c(Link);
  MP_variable x(S,D);
  MP_constraint supply(S), demand(D);

  c(Link) = f * d(Link) / 1000;

  supply(S) = sum( Link(S,D), x(Link) ) <= SUPPLY(S);
  demand(D) = sum( Link(S,D), x(Link) ) >= DEMAND(D);

  minimize( sum(Link, c(Link) * x(Link)) );

  x.display("Optimal transportation plan");
}
```

# MP\_expression



- Explicitly name a linear expression without adding rows or columns to the generated instance.
- Alternative to accounting variables

```
MP_expression production(Products);
```

```
production(p) = sum(T(t), xi(p,t) + xo(p,t));
```

```
Min_prod(p) = production(p) >= MIN_PROD(p);
```

```
E_limit(q) =  
    sum(Products(p), E(p,q)*production(p)) <= LIMIT(q);
```

# minimize\_max



- Higher level modelling extensions
- Modelling objects as parameters

```
void MP_model::minimize_max(MP_set &s, const MP_expression &obj) {  
    MP_variable v;  
    MP_constraint c(s);  
    add(c);  
    c(s) = v() >= obj;  
    minimize(v());  
}  
.....  
minimize_max( J, s(J)+D(J) );
```

# Problem modification I



Routines to change

- the objective function
- lower and upper bounds on variables
- right hand side and sense of constraints
- variable types

or add or delete

- columns (variables)
- rows (constraints)

without regeneration of the problem.

# Problem modification II



```
for (int j=0; j<numDistributionCenters; j++)
    subproblem->setRowBounds(selling(j), received.level(j), received.level(j));
double objsub = 0.0;
for (int s=0; s<numScenarios; s++) {
    for (int j=0; j<numDistributionCenters; j++)
        subproblem->setRowBounds( selmax(j), Demand[s][j], Demand[s][j]);
    subproblem->resolve();
    objsub += prob[s]*subproblem->getObjValue();
    for (int j=0; j<numDistributionCenters; j++) {
        cutCnst += prob[s]*selmax.price(j)*Demand[j][s];
        coeff(j) += prob[s]*selling.price(j);
    }
}
upperbound = min(upperbound, objmaster + objsub);
if ((upperbound-lowerbound) < 0.01*(1+fabs(lowerbound))) break;
masterproblem.addRow( theta() >= cutCnst + sum(j, coeff(j)*received(j)) )
masterproblem->resolve();
lowerbound = masterproblem->getObjValue();
objmaster = lowerbound - theta.level();
```



# Cyclic index sets



```
enum {t12pm_6am, t6am_9am, t9am_3pm, t3pm_6pm, t6pm_12pm, numT};
```

```
MP_set T(numT); T.cyclic();
```

```
MP_set G(numG);
```

```
MP_integer_variable n(G,T); // number of generators in use
```

```
MP_variable s(G,T); // number of generators started up
```

```
MP_constraint st(G,T); // start_up definition
```

```
st(G,T) = s(G,T) >= n(G,T) - n(G,T-1);
```

(full model in magic.cpp)

# Implementation



- keep track of correspondence between individual variables (/constraints) and its column (/row) index in the generated problem instance
- expressions evaluate to abstract syntax trees

```
supply(S) = sum( Link(S,D), x(Link) ) <= SUPPLY(S);  
demand(D) = sum( Link(S,D), x(Link) ) >= DEMAND(D);
```

- generation takes place in the leaf nodes, information from upper level nodes is passed recursively down in the tree
- expressions are reference counted to avoid memory leaks

# Miscellaneous



- Selfcontained.
  - Required Coin-OR projects (CoinUtils, Cgl, Clp, Cbc, Osi, MSVisualStudio, BuildTools) included.
- Solver independent.
  - Uses OSI (Optimization Solver Interface) from Coin-OR (Clp, Cbc, CPLEX, DyLP, FortMP, GLPK, MOSEK, OSL, SoPlex, SYMPHONY, Vol, XPRESS-MP)
- Windows / MS VisualStudio supported.

# Documentation



1. Doxygen Documentation. (Philip Walton)
2. 'User Documentation' (Christian Woellenstein)
3. Small FAQ (pose questions to the mailing list)
4. Examples (currently 18 different)

# Open source



Initiation of optimization projects:

- One less tool to buy/license

Long term maintenance:

- Compiler release versioning – do we have the right run time library for this platform?
- Operating system release library versions – do we have the build for this machine?
- Licensing hassles (revisit the code because of licensing changes). Must keep feeding the maintenance cost.

Development:

- Users are able to contribute - and they do.

# Plans



- Symbolic model checks

```
MP_set I(5), J(2);  
MP_variable x(I,J);  
MP_constraint c;  
c() = sum(I*J, x(J,I)) <= 100
```

generates this (without warnings):

```
x(0,0) + x(0,1) + x(1,0) + x(1,1) <= 100
```

- Non-linear programming modelling, etc. Possible, but...
- Stochastic programming modelling.
  - In the pipeline...

# SP extensions



## SP with current FLOPC++

- Benders decomposition (stochBenders.cpp)
- Recursive formulation (stampI.cpp)

## Work in progres (with Alan King)

- `MP_stage T(numStages) ;`
- `MP_stochastic_data Return(T, INSTR) ;`
- Uses Coin-Smi (Stochastic Modelling interface)
- Generate core model
- Add scenarios

# SP example



```
MP_set INSTR(numINSTR);
MP_stage T(numStages);
MP_stochastic_data Return(T,INSTR);

MP_variable Buy(T,INSTR), Shortage, Overage;
MP_constraint InvestAll, ReinvestAll(T), Goal;

InvestAll() = sum(INSTR, Buy(0,INSTR)) == initial_wealth;

ReinvestAll(T+1) = sum(INSTR, Buy(T,INSTR) * Return(T,INSTR))
                 == sum(INSTR, Buy(T+1,INSTR));

Goal() = sum(INSTR, Buy(T.last(),INSTR) * Return(T.last(),INSTR))
        == goal - Shortage() + Overage();

maximize( Overage() - 4*Shortage() );
```



# Conclusion



**Advantages** in addition to traditional modelling languages.

- Lightweight
- Open source
- Multi paradigm
- Fast problem generation
- Seamless integration with applications
- Efficient and powerful for customized algorithms (decomposition, column generation, cut generation)
- Modelling objects are first class C++ types. This allows higher level modelling extensions (such as the minimize max objective) to be implemented.

**Disadvantages ?**