

New AMPL Interfaces for Enhanced Optimization Model Development and Deployment



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Track 11, Monday 1:50-2:40, *Software Tutorials*

Outline

Deploying models

- ❖ Scripting
 - * Internal modeling/programming language
- ❖ **AMPL API** (Application Programming Interfaces)
 - * External programming language support
 - * General-purpose languages: C++, Java, .NET, Python
 - * Analytics languages: MATLAB, R

Developing models

- ❖ More natural formulations
 - * Logical conditions
 - * Quadratic constraints
- ❖ **AMPL IDE** (Integrated Development Environment)
 - * Unified editor & command processor
 - * Built on the Eclipse platform

Introductory Example

Multicommodity transportation . . .

- ❖ Products available at factories
- ❖ Products needed at stores
- ❖ Plan shipments at lowest cost

. . . with practical restrictions

- ❖ Cost has fixed and variable parts
- ❖ Shipments cannot be too small
- ❖ Factories cannot serve too many stores

Multicommodity Transportation

Given

- O Set of origins (factories)
- D Set of destinations (stores)
- P Set of products

and

- a_{ip} Amount available, for each $i \in O$ and $p \in P$
- b_{jp} Amount required, for each $j \in D$ and $p \in P$
- l_{ij} Limit on total shipments, for each $i \in O$ and $j \in D$
- c_{ijp} Shipping cost per unit, for each $i \in O, j \in D, p \in P$
- d_{ij} Fixed cost for shipping any amount from $i \in O$ to $j \in D$
- s Minimum total size of any shipment
- n Maximum number of destinations served by any origin

Multicommodity Transportation

Mathematical Formulation

Determine

X_{ijp} Amount of each $p \in P$ to be shipped from $i \in O$ to $j \in D$

Y_{ij} 1 if any product is shipped from $i \in O$ to $j \in D$
0 otherwise

to minimize

$$\sum_{i \in O} \sum_{j \in D} \sum_{p \in P} c_{ijp} X_{ijp} + \sum_{i \in O} \sum_{j \in D} d_{ij} Y_{ij}$$

Total variable cost plus total fixed cost

Multicommodity Transportation

Mathematical Formulation

Subject to

$$\sum_{j \in D} X_{ijp} \leq a_{ip} \quad \text{for all } i \in O, p \in P$$

Total shipments of product p out of origin i
must not exceed availability

$$\sum_{i \in O} X_{ijp} = b_{jp} \quad \text{for all } j \in D, p \in P$$

Total shipments of product p into destination j
must satisfy requirements

Mathematical Formulation

Subject to

$$\sum_{p \in P} X_{ijp} \leq l_{ij} Y_{ij} \quad \text{for all } i \in O, j \in D$$

When there are shipments from origin i to destination j , the total may not exceed the limit, and Y_{ij} must be 1

$$\sum_{p \in P} X_{ijp} \geq s Y_{ij} \quad \text{for all } i \in O, j \in D$$

When there are shipments from origin i to destination j , the total amount of shipments must be at least s

$$\sum_{j \in D} Y_{ij} \leq n \quad \text{for all } i \in O$$

Number of destinations served by origin i must be at most n

AMPL Formulation

Symbolic data

```
set ORIG;    # origins
set DEST;    # destinations
set PROD;    # products

param supply {ORIG,PROD} >= 0; # availabilities at origins
param demand {DEST,PROD} >= 0; # requirements at destinations
param limit  {ORIG,DEST} >= 0; # capacities of links

param vcost  {ORIG,DEST,PROD} >= 0; # variable shipment cost
param fcost  {ORIG,DEST} > 0;      # fixed usage cost

param minload >= 0;                # minimum shipment size
param maxserve integer > 0;       # maximum destinations served
```


AMPL Formulation

Symbolic model: variables and objective

```
var Trans {ORIG,DEST,PROD} >= 0;    # actual units to be shipped
var Use {ORIG, DEST} binary;        # 1 if link used, 0 otherwise

minimize Total_Cost:
    sum {i in ORIG, j in DEST, p in PROD} vcost[i,j,p] * Trans[i,j,p]
+ sum {i in ORIG, j in DEST} fcost[i,j] * Use[i,j];
```

$$\sum_{i \in O} \sum_{j \in D} \sum_{p \in P} c_{ijp} X_{ijp} + \sum_{i \in O} \sum_{j \in D} d_{ij} Y_{ij}$$

Multicommodity Transportation

AMPL Formulation

Symbolic model: constraint

```
subject to Supply {i in ORIG, p in PROD}:  
    sum {j in DEST} Trans[i,j,p] <= supply[i,p];
```

$$\sum_{j \in D} X_{ijp} \leq a_{ip}, \text{ for all } i \in O, p \in P$$

AMPL Formulation

Symbolic model: constraints

```
subject to Supply {i in ORIG, p in PROD}:  
    sum {j in DEST} Trans[i,j,p] <= supply[i,p];  
  
subject to Demand {j in DEST, p in PROD}:  
    sum {i in ORIG} Trans[i,j,p] = demand[j,p];  
  
subject to Multi {i in ORIG, j in DEST}:  
    sum {p in PROD} Trans[i,j,p] <= limit[i,j] * Use[i,j];  
  
subject to Min_Ship {i in ORIG, j in DEST}:  
    sum {p in PROD} Trans[i,j,p] >= minload * Use[i,j];  
  
subject to Max_Serve {i in ORIG}:  
    sum {j in DEST} Use[i,j] <= maxserve;
```

AMPL Formulation

Explicit data independent of symbolic model

```
set ORIG := GARY CLEV PITT ;
set DEST := FRA DET LAN WIN STL FRE LAF ;
set PROD := bands coils plate ;

param supply (tr):  GARY  CLEV  PITT :=
                    bands  400   700   800
                    coils  800  1600  1800
                    plate  200   300   300 ;

param demand (tr):
                    FRA  DET  LAN  WIN  STL  FRE  LAF :=
bands  300  300  100  75  650  225  250
coils  500  750  400  250  950  850  500
plate  100  100   0   50  200  100  250 ;

param limit default 625 ;

param minload := 375 ;
param maxserve := 5 ;
```

AMPL Formulation

Explicit data (continued)

```
param vcost :=
  [*,*,bands]: FRA DET LAN WIN STL FRE LAF :=
    GARY 30 10 8 10 11 71 6
    CLEV 22 7 10 7 21 82 13
    PITT 19 11 12 10 25 83 15
  [*,*,coils]: FRA DET LAN WIN STL FRE LAF :=
    GARY 39 14 11 14 16 82 8
    CLEV 27 9 12 9 26 95 17
    PITT 24 14 17 13 28 99 20
  [*,*,plate]: FRA DET LAN WIN STL FRE LAF :=
    GARY 41 15 12 16 17 86 8
    CLEV 29 9 13 9 28 99 18
    PITT 26 14 17 13 31 104 20 ;
param fcost: FRA DET LAN WIN STL FRE LAF :=
  GARY 3000 1200 1200 1200 2500 3500 2500
  CLEV 2000 1000 1500 1200 2500 3000 2200
  PITT 2000 1200 1500 1500 2500 3500 2200 ;
```

Multicommodity Transportation

AMPL Solution

Model + data = problem instance to be solved

```
ampl: model multmipG.mod;
ampl: data multmipG.dat;
ampl: option solver gurobi;
ampl: solve;
Gurobi 5.6.0: optimal solution; objective 235625
293 simplex iterations
28 branch-and-cut nodes
ampl: display Use;
Use [*,*]
:      DET FRA FRE LAF LAN STL WIN  :=
CLEV   1   1   1   0   1   1   0
GARY   0   0   0   1   0   1   1
PITT   1   1   1   1   0   1   0
;
```

Multicommodity Transportation

AMPL Solution

Solver choice independent of model and data

```
AMPL: model multmipG.mod;
AMPL: data multmipG.dat;
AMPL: option solver cplex;
AMPL: solve;
CPLEX 12.6.0.0: optimal integer solution; objective 235625
136 MIP simplex iterations
0 branch-and-bound nodes
AMPL: display Use;
Use [*,*]
:      DET FRA FRE LAF LAN STL WIN  :=
CLEV   1   1   1   0   1   1   0
GARY   0   0   0   1   0   1   1
PITT   1   1   1   1   0   1   0
;
```

Multicommodity Transportation

AMPL Solution

Examine results

```
AMPL: display {i in ORIG, j in DEST}
AMPL?   sum {p in PROD} Trans[i,j,p] / limit[i,j];

:      DET    FRA    FRE    LAF    LAN    STL    WIN    :=
CLEV   1      0.6    0.88   0     0.8    0.88   0
GARY   0      0      0     0.64   0     1      0.6
PITT   0.84    0.84   1     0.96   0     1      0
;

AMPL: display Max_Serve.body;

CLEV   5
GARY   3
PITT   5
;

AMPL: display TotalCost,
AMPL?   sum {i in ORIG, j in DEST} fcost[i,j] * Use[i,j];

TotalCost = 235625
sum {i in ORIG, j in DEST} fcost[i,j]*Use[i,j] = 27600
```


Multicommodity Transportation

AMPL “Sparse” Network

Indexed over sets of pairs and triples

```
set ORIG;    # origins
set DEST;    # destinations
set PROD;    # products

set SHIP within {ORIG,DEST,PROD};
           # (i,j,p) in SHIP ==> can ship p from i to j
set LINK = setof {(i,j,p) in SHIP} (i,j);
           # (i,j) in LINK ==> can ship some products from i to j

.....

var Trans {SHIP} >= 0;    # actual units to be shipped
var Use {LINK} binary;    # 1 if link used, 0 otherwise

minimize Total_Cost:
    sum {(i,j,p) in SHIP} vcost[i,j,p] * Trans[i,j,p]
+ sum {(i,j) in LINK} fcost[i,j] * Use[i,j];
```

Scripting

Incorporate programming concepts . . .

- ❖ Loops of various kinds
- ❖ If-then and If-then-else conditionals
- ❖ Assignments

. . . using modeling language syntax

- ❖ Same algebraic expressions
- ❖ Same set indexing expressions
- ❖ All commands from interactive mode

1: Parametric Analysis

Try different limits on destinations served

- ❖ Reduce parameter `maxserve` and re-solve
 - * until there is no feasible solution
- ❖ Display results
 - * parameter value
 - * numbers of destinations actually served

Try different supplies of plate at Gary

- ❖ Increase parameter `supply['GARY', 'plate']` and re-solve
 - * until dual is zero (constraint is slack)
- ❖ Record results
 - * distinct dual values
 - * corresponding objective values

... display results at the end

Parametric Analysis *on limits*

Script to test sensitivity to serve limit

```
model multmipG.mod;
data multmipG.dat;

option solver gurobi;

for {m in 7..1 by -1} {
  let maxserve := m;
  solve;
  if solve_result = 'infeasible' then break;
  display maxserve, Max_Serve.body;
}
```

```
subject to Max_Serve {i in ORIG}:
  sum {j in DEST} Use[i,j] <= maxserve;
```

Parametric Analysis *on limits*

Run showing sensitivity to serve limit

```
ampl: include multmipServ.run;

Gurobi 5.6.0: optimal solution; objective 233150
maxserve = 7
CLEV 5   GARY 3   PITT 6

Gurobi 5.6.0: optimal solution; objective 233150
maxserve = 6
CLEV 5   GARY 3   PITT 6

Gurobi 5.6.0: optimal solution; objective 235625
maxserve = 5
CLEV 5   GARY 3   PITT 5

Gurobi 5.6.0: infeasible
```

Parametric Analysis on *supplies*

Script to test sensitivity to plate supply at GARY

```
set SUPPLY default {};  
param sup_obj {SUPPLY};  
param sup_dual {SUPPLY};  
  
let supply['GARY','plate'] := 200;  
param sup_step = 10;  
param previous_dual default -Infinity;  
repeat while previous_dual < 0 {  
  solve;  
  if Supply['GARY','plate'].dual > previous_dual then {  
    let SUPPLY := SUPPLY union {supply['GARY','plate']};  
    let sup_obj[supply['GARY','plate']] := Total_Cost;  
    let sup_dual[supply['GARY','plate']] := Supply['GARY','plate'].dual;  
    let previous_dual := Supply['GARY','plate'].dual;  
  }  
  
  let supply['GARY','plate'] := supply['GARY','plate'] + supply_step;  
}
```

Parametric Analysis on *supplies*

Run showing sensitivity to plate supply at GARY

```
ampl: include multmipSupply.run;

ampl: display sup_obj, sup_dual;

:      sup_obj    sup_dual    :=
200    223504     -13
380    221171     -11.52
460    220260     -10.52
510    219754     -8.52
560    219413     0
;
```

Parametric: Observations

Results of solve can be tested

- ❖ Check whether problem is no longer feasible
 - * `if solve_result = 'infeasible' then break;`

Parameters are true objects

- ❖ Assign new value to param `supply`
 - * `let supply['GARY','plate'] := supply['GARY','plate'] + supply_step;`
- ❖ Problem instance changes accordingly

Sets are true data

- ❖ Assign new value to set `SUPPLY`
 - * `let SUPPLY := SUPPLY union {supply['GARY','plate']};`
- ❖ All indexed entities change accordingly

2a: Solution Generation *via Cuts*

Same multicommodity transportation model

Generate n best solutions using different routes

- ❖ Display routes used by each solution

Solutions *via Cuts*

Script

```
param nSols default 0;
param maxSols = 3;

model multmipG.mod;
data multmipG.dat;

set USED {1..nSols} within {ORIG,DEST};

subject to exclude {k in 1..nSols}:
    sum {(i,j) in USED[k]} (1-Use[i,j]) +
    sum {(i,j) in {ORIG,DEST} diff USED[k]} Use[i,j] >= 1;

repeat {
    solve;
    display Use;
    let nSols := nSols + 1;
    let USED[nSols] := {i in ORIG, j in DEST: Use[i,j] > .5};
} until nSols = maxSols;
```

Solutions *via Cuts*

Run showing 3 best solutions

```
ampl: include multmipBestA.run;
Gurobi 5.6.0: optimal solution; objective 235625
:      DET FRA FRE LAF LAN STL WIN      :=
CLEV   1   1   1   0   1   1   0
GARY   0   0   0   1   0   1   1
PITT   1   1   1   1   0   1   0 ;
Gurobi 5.6.0: optimal solution; objective 237125
:      DET FRA FRE LAF LAN STL WIN      :=
CLEV   1   1   1   1   0   1   0
GARY   0   0   0   1   0   1   1
PITT   1   1   1   0   1   1   0 ;
Gurobi 5.6.0: optimal solution; objective 238225
:      DET FRA FRE LAF LAN STL WIN      :=
CLEV   1   0   1   0   1   1   1
GARY   0   1   0   1   0   1   0
PITT   1   1   1   1   0   1   0 ;
```

Solutions *via Cuts*: Observations

Same expressions describe sets and indexing

- ❖ Index a summation
 - * ... `sum {(i,j) in {ORIG,DEST} diff USED[k]} Use[i,j] >= 1;`
- ❖ Assign a value to a set
 - * `let USED[nSols] := {i in ORIG, j in DEST: Use[i,j] > .5};`

New cuts defined automatically

- ❖ Index cuts over a set
 - * `subject to exclude {k in 1..nSols}: ...`
- ❖ Add a cut by expanding the set
 - * `let nSols := nSols + 1;`

2b: Solution Generation *via Solver*

Same model

Ask solver to return multiple solutions

- ❖ Set options
- ❖ Get all results from one “solve”
- ❖ Retrieve and display each solution

Solutions *via Solver*

Script

```
option solver cplex;  
option cplex_options "poolstub=multmip poolcapacity=3 \  
    populate=1 poolintensity=4 poolreplace=1";  
  
solve;  
  
for {i in 1..Current.npool} {  
    solution ("multmip" & i & ".sol");  
    display Use;  
}
```

Solutions *via Solver*

Results

```
ampl: include multmipBestB.run;
CPLEX 12.6.0.0: poolstub=multmip
poolcapacity=3
populate=1
poolintensity=4
poolreplace=1

CPLEX 12.6.0.0: optimal integer solution; objective 235625
742 MIP simplex iterations
56 branch-and-bound nodes

Wrote 3 solutions in solution pool
to files multmip1.sol ... multmip3.sol.

Suffix npool OUT;
```

Solutions *via Solver*

Results (continued)

Solution pool member 1 (of 3); objective 235625

```
:      DET FRA FRE LAF LAN STL WIN  :=
CLEV   1   1   1   0   1   1   0
GARY   0   0   0   1   0   1   1
PITT   1   1   1   1   0   1   0 ;
```

Solution pool member 2 (of 3); objective 238225

```
:      DET FRA FRE LAF LAN STL WIN  :=
CLEV   1   0   1   0   1   1   1
GARY   0   1   0   1   0   1   0
PITT   1   1   1   1   0   1   0 ;
```

Solution pool member 3 (of 3); objective 237125

```
:      DET FRA FRE LAF LAN STL WIN  :=
CLEV   1   1   1   1   0   1   0
GARY   0   0   0   1   0   1   1
PITT   1   1   1   0   1   1   0 ;
```


Solutions *via Solver*: Observations

Filenames can be formed dynamically

- ❖ Write a (string expression)
- ❖ Numbers are automatically converted
 - * `solution ("multmip" & i & ".sol");`

Scripting

General Observations

Scripts in practice

- ❖ Large and complicated
 - * Multiple files
 - * Hundreds of statements
 - * Millions of statements executed
- ❖ Run within broader applications

Prospective improvements

- ❖ Faster loops
- ❖ True script functions
 - * Arguments and return values
 - * Local sets & parameters
 - * Callback functions

But . . .

Scripting

Limitations

Performance

- ❖ Interpreted language
- ❖ General set & data structures

Power

- ❖ Not a complete programming language
 - * Specific to optimization models
- ❖ Not object-oriented

So . . .

AMPL API

Application Programming Interface

- ❖ General-purpose languages: C++, Java, .NET, Python
- ❖ Analytics languages: MATLAB, R

Facilitates use of AMPL for

- ❖ Complex algorithmic schemes
- ❖ Embedding in other applications
- ❖ Deployment of models

Example: Efficient Frontier

Compute a series of “nondominated” portfolios

- ❖ Model in AMPL
- ❖ Data & logic from a programming language
 - * Java
 - * MATLAB

Key to examples

- ❖ AMPL entities
- ❖ Java/MATLAB objects
- ❖ Java/MATLAB methods for working with AMPL

AMPL Model Files

Sets, parameters, variables

```
set stockall;           # All stocks in the universe of assets
set stockrun within stockall; # Stocks used for a given run
set stockopall within stockrun; # Stocks that had weight > 0.5

param ncard default 10; # Maximum cardinality of portfolio
param averret {stockall}; # Average return of each stock
param covar {stockall,stockall}; # Covariance matrix
param cutoffl default 0.0001; # Weight > low cutoff ==> stock out
param cutofffh {stockall} default 1; # Weight > high cutoff ==> stock in
param targetret default 0; # Target return for efficient frontier

var weights {stockrun} >= 0; # Weight of each stock in current run
var ifstock {stockrun} binary; # = 1 iff a stock is in
var portret >= targetret; # Portfolio return
```

AMPL Model Files

Objective, constraints

```
minimize cst: # Overall risk
    sum {s in stockrun,s1 in stockrun} weights[s] * covar[s,s1] * weights[s1];

subject to invest: # Weights must total 1
    sum{s in stockrun} weights[s]=1;

subject to defret: # Returns must equal specified level
    sum{s in stockrun} averret[s] * weights[s] = portret;

subject to lowlnk {s in stockrun}: # Weight >= low cutoff if stock is in
    weights[s] >= cutoffl * ifstock[s];

subject to uplink {s in stockrun}: # Weight <= high cutoff if stock is in
    weights[s] <= cutoffh[s] * ifstock[s];

subject to fixing {s in stockopall}: # Specified stocks must be in
    ifstock[s] = 1;

subject to carda: # Limit on number of stocks in
    sum {s in stockrun} ifstock[s] <= ncard;
```

AMPL API Example

AMPL Model Files

Data table definitions

```
param data_dir symbolic;  
table assetstable IN (data_dir & "/assetsReturns.bit"):  
    stockall <- [stockall], averret;  
table astrets IN (data_dir & "/covar.bit"):  
    [Asset, stockall], covar;
```


AMPL API Example

Java Program

AMPL packages

```
package com.ampl.examples;  
  
import com.ampl.AMPL;  
import com.ampl.Objective;  
import com.ampl.Parameter;  
import com.ampl.Variable;  
  
import java.io.IOException;
```

AMPL API Example

Java Program

Main class & method

```
public class EfficientFrontier {
    public static void main(String[] args) {
        int steps = 10;
        String modelDirectory = args[0];
        // initialize AMPL object
        AMPL ampl = new AMPL();
        try {
            // DETAILS OF THE PROGRAM
        } catch (IOException e) {
            System.out.println ("Model file not found.");
        } finally {
            ampl.close();
        }
    }
    // definitions of auxiliary methods
}
```

AMPL API Example

Java Program

Setup

```
// set AMPL options
ampl.setBoolOption("reset_initial_guesses", true);
ampl.setBoolOption("send_statuses", false);

// load AMPL model from file
ampl.read(modelDirectory + "/qpmv.mod");
ampl.read(modelDirectory + "/qpmvbit.run");

// pass tables directory to AMPL, then read tables
ampl.getParameter("data_dir").set(modelDirectory);

ampl.readTable("assetstable");
ampl.readTable("astrets");
```

AMPL API Example

Java Program

Initial solve

```
// associate Java variables with AMPL entities
Variable portfolioReturn = ampl.getVariable("portret");
Parameter averageReturn = ampl.getParameter("averret");
Parameter targetReturn = ampl.getParameter("targetret");
Objective deviation = ampl.getObjective("cst");

// run AMPL commands directly
ampl.eval("option solver gurobi;");
ampl.eval("let stockopall := {};");
ampl.eval("let stockrun := stockall;");

// relax integrality
ampl.setBoolOption("relax_integrality", true);

// solve
ampl.solve();
```

Java Program

Solve loop

```
// calibrate efficient frontier range
double minret = portfolioReturn.value();
double maxret = fMax(averageReturn.getValues().getColumnAsDoubles("val"));
double stepsize = (maxret - minret) / steps;

// initialize arrays to hold results
double[] returns = new double[steps];
double[] deviations = new double[steps];

for (int i = 0; i < steps; i++) {
    // SOLVE LOOP
}

// Display efficient frontier points
System.out.format("%-8s  %-8s\n", "RETURN", "DEVIATION");
for (int i = 0; i < returns.length; i++)
    System.out.format("%-6f  %-6f\n", returns[i], deviations[i]);
```

Java Program

Inside the solve loop (1)

```
for (int i = 0; i < steps; i++) {
    System.out.format
        ("Solving for return = %f%n", maxret - (i-1)*stepsize);
    // set target return to desired point
    targetReturn.setValues(maxret - (i-1) * stepsize);
    ampl.eval("let stockopall := {};");
    ampl.eval("let stockrun := stockall;");
    // relax integrality and solve
    ampl.setBoolOption("relax_integrality", true);
    ampl.solve();
    System.out.format("QP result = %f %n", deviation.value());
}
```

Java Program

Inside the solve loop (2)

```
// adjust included stocks
AMPL.eval("let stockrun := {i in stockrun: weights[i] > 0};");
AMPL.eval("let stockopall := {i in stockrun: weights[i] > 0.5};");

// restore integrality
AMPL.setBoolOption("relax_integrality", false);
AMPL.solve();

System.out.format("QMIP result = %f%n", deviation.value());

// save current frontier point
returns[i] = maxret - (i-1)*stepsize;
deviations[i] = deviation.value();
}
```

MATLAB Program

Setup

```
steps = 10;
modelDirectory = '/amplapi/examples/';

% initialize AMPL object
AMPL = initAMPL;

% set AMPL options
AMPL.setBoolOption("reset_initial_guesses", true);
AMPL.setBoolOption("send_statuses", false);

% load AMPL model from file
AMPL.read([modelDirectory 'qpmv.mod'])
AMPL.read([modelDirectory 'qpmvbit.run'])

// pass tables directory to AMPL, then read tables
AMPL.getParameter('data_dir').set(modelDirectory);

AMPL.readTable('assetstable');
AMPL.readTable('astrets');
```


AMPL API Example

MATLAB Program

Initial solve

```
% associate Java variables with AMPL entities
portfolioReturn = ampl.getVariable('portret');
averageReturn = ampl.getParameter('averret');
targetReturn = ampl.getParameter('targetret');
deviation = ampl.getObjective('cst');

% run AMPL commands directly
ampl.eval('option solver gurobi;');
ampl.eval('let stockopall := {};');
ampl.eval('let stockrun := stockall;');

% relax integrality
ampl.setBoolOption('relax_integrality', true);

% solve
ampl.solve();
```

AMPL API Example

MATLAB Program

Solve loop

```
% calibrate efficient frontier range
minret = portfolioReturn.value();
maxret = max(averageReturn.getValues());
stepsize = (maxret-minret)/steps;

% initialize arrays to hold results
returns = zeros(steps, 1);
deviations = zeros(steps, 1);

for i=1:steps
    % SOLVE LOOP
End

% Plot efficient frontier points
plot(returns, deviations)
```

AMPL API Example

MATLAB Program

Inside the solve loop (1)

```
for i=1:steps
    fprintf('Solving for return = %f\n', maxret - (i-1)*stepsize)
    % set target return to desired point
    targetReturn.setValues(maxret - (i-1)*stepsize);
    ampl.eval('let stockopall:={ };');
    ampl.eval('let stockrun:=stockall;');
    % relax integrality and solve
    ampl.setBoolOption('relax_integrality', 1);
    ampl.solve();
    fprintf('QP result = %f ', deviation.value())
```

AMPL API Example

MATLAB Program

Inside the solve loop (2)

```
% adjust included stocks
AMPL.eval('let stockrun := {i in stockrun: weights[i]>0};');
AMPL.eval('let stockopall := {i in stockrun: weights[i]>0.5};');

% restore integrality
AMPL.setBoolOption('relax_integrality', 0);
AMPL.solve();

fprintf(' QMIP result = %f\n ', deviation.value());

% save current frontier point
returns(i) = maxret - (i-1)*stepsize;
deviations(i) = deviation.value();
end
```

Data Transfer

Process

- ❖ Define symbolic sets & parameters in AMPL model
- ❖ Create corresponding objects in program
- ❖ Transfer data using API methods
 - * Program to AMPL
 - * AMPL to program

Methods for transfer between . . .

- ❖ Scalar values
- ❖ Collections of values
 - * AMPL indexed expressions
 - * Java arrays, MATLAB matrices
- ❖ Relational tables
 - * AMPL “table” structures
 - * API DataFrame objects in Java, MATLAB

Deployment Alternatives

Stand-alone: Give (temporary) control to AMPL

- ❖ Write needed files
- ❖ Invoke AMPL to run some scripts
- ❖ Read the files that AMPL leaves on exit

API: Interact with AMPL

- ❖ Execute AMPL statements individually
- ❖ Read model, data, script files when convenient
- ❖ Exchange data tables directly with AMPL
 - * populate sets & parameters
 - * invoke any available solver
 - * extract values of variables & result expressions

. . . all embedded within your program's logic

Planned Availability

Initial languages: Java, MATLAB

- ❖ Beta test
 - * April 2014
 - * *Seeking beta testers now*
- ❖ Release
 - * Summer 2014
 - * Available with all AMPL distributions

More languages to follow

- ❖ C++, C# (.NET), Python, R

Development details

- ❖ Partnership with OptiRisk Systems
- ❖ Long-term development & maintenance by AMPL

More Natural Formulations

Logical conditions

- ❖ Implications (\implies)
- ❖ Disjunctions (or)

Convex quadratics

- ❖ Objectives and constraints
- ❖ Elliptic and conic

Multicommodity Revisited

Minimum-shipment constraints

- ❖ From each origin to each destination, *either* ship nothing *or* ship at least minload units

Conventional linear mixed-integer formulation

```
var Trans {ORIG,DEST,PROD} >= 0;
var Use {ORIG, DEST} binary;
.....
subject to Multi {i in ORIG, j in DEST}:
    sum {p in PROD} Trans[i,j,p] <= limit[i,j] * Use[i,j];
subject to Min_Ship {i in ORIG, j in DEST}:
    sum {p in PROD} Trans[i,j,p] >= minload * Use[i,j];
```

Multi-Commodity

Zero-One Alternatives

Mixed-integer formulation using implications

```
subject to Multi_Min_Ship {i in ORIG, j in DEST}:  
  Use[i,j] = 1 ==>  
    minload <= sum {p in PROD} Trans[i,j,p] <= limit[i,j]  
    else sum {p in PROD} Trans[i,j,p] = 0;
```

Solved directly by CPLEX

```
ampl: model multmipImpl.mod;  
ampl: data multmipG.dat;  
ampl: option solver cplex;  
ampl: solve;  
CPLEX 12.5.1.0: optimal integer solution; objective 235625  
176 MIP simplex iterations  
0 branch-and-bound nodes
```

Multi-Commodity

Non-Zero-One Alternatives

Disjunctive constraint

```
subject to Multi_Min_Ship {i in ORIG, j in DEST}:  
    sum {p in PROD} Trans[i,j,p] = 0 or  
    minload <= sum {p in PROD} Trans[i,j,p] <= limit[i,j];
```

Solved by CPLEX?

```
AMPL: model multmipDisj.mod;  
AMPL: data multmipG.dat;  
AMPL: solve;  
CPLEX 12.5.1.0: logical constraint not indicator constraint.  
AMPL: option solver ilogcp;  
AMPL: option ilogcp_options 'optimizer cplex';  
AMPL: solve;  
ilogcp 12.5.0: optimal solution  
0 nodes, 175 iterations, objective 235625
```

More Natural Modeling

Logical Conditions: Current

Expressions recognized by AMPL

- ❖ Disjunctions (or), implications (\implies)
- ❖ Counting expressions (count),
Counting constraints (atleast, atmost)
- ❖ Aggregate constraints (alldiff, numberof)

Solvers supported

- ❖ IBM CPLEX mixed-integer programming solver
 - * Applied directly
 - * Applied after automatic conversion to MIP
- ❖ Constraint programming solvers
 - * IBM ILOG CP
 - * Gecode
 - * JaCoP

More Natural Modeling

Logical Conditions: Planned

What the AMPL-solver interface will do

- ❖ Recognize transformable “not linear” expressions
 - * Logical operators
 - * Piecewise operators: abs, min, max
- ❖ Automatically transform to LPs or MILPs

New forms to be recognized

- ❖ Object-valued variables
 - * `var JobForSlot {1..nSl+1} in JOBS;`
- ❖ Variables in subscripts
 - * `minimize TotalCost:`
`sum {k in 1..nSl} setupCost [JobForSlot [k], JobForSlot [k+1]] + ...`
- ❖ Set membership constraints
 - * `subject to SeqRestrictions {k in 1..nSl}`
`(JobForSlot [k], JobForSlot [k+1]) in ALLOWED;`

More Natural Modeling

Convex Quadratics: Current

Problem types

- ❖ Elliptic: quadratic programs (QPs)
- ❖ Conic: second-order cone programs (SOCPs)

What the AMPL-solver interface does

- ❖ Recognize quadratic objectives & constraints
- ❖ Multiply out products of linear terms
- ❖ Send linear & quadratic coefficient lists to solver

What the solver does

- ❖ Detect elliptic forms numerically
- ❖ Detect conic forms by structural analysis

More Natural Modeling

Convex Quadratics: **Planned**

What the AMPL-solver interface will do

- ❖ Recognize nonquadratic SOCP-equivalent problems
- ❖ Automatically transform to SOCPs recognizable by solvers

Forms to be recognized

- ❖ Sum of norms
- ❖ Sum of squares divided by linear
- ❖ Generalized geometric means
- ❖ Generalized p -norms
- ❖ log-Chebyshev objectives

*... combinations by sum, max, positive multiple
(where possible)*

AMPL IDE

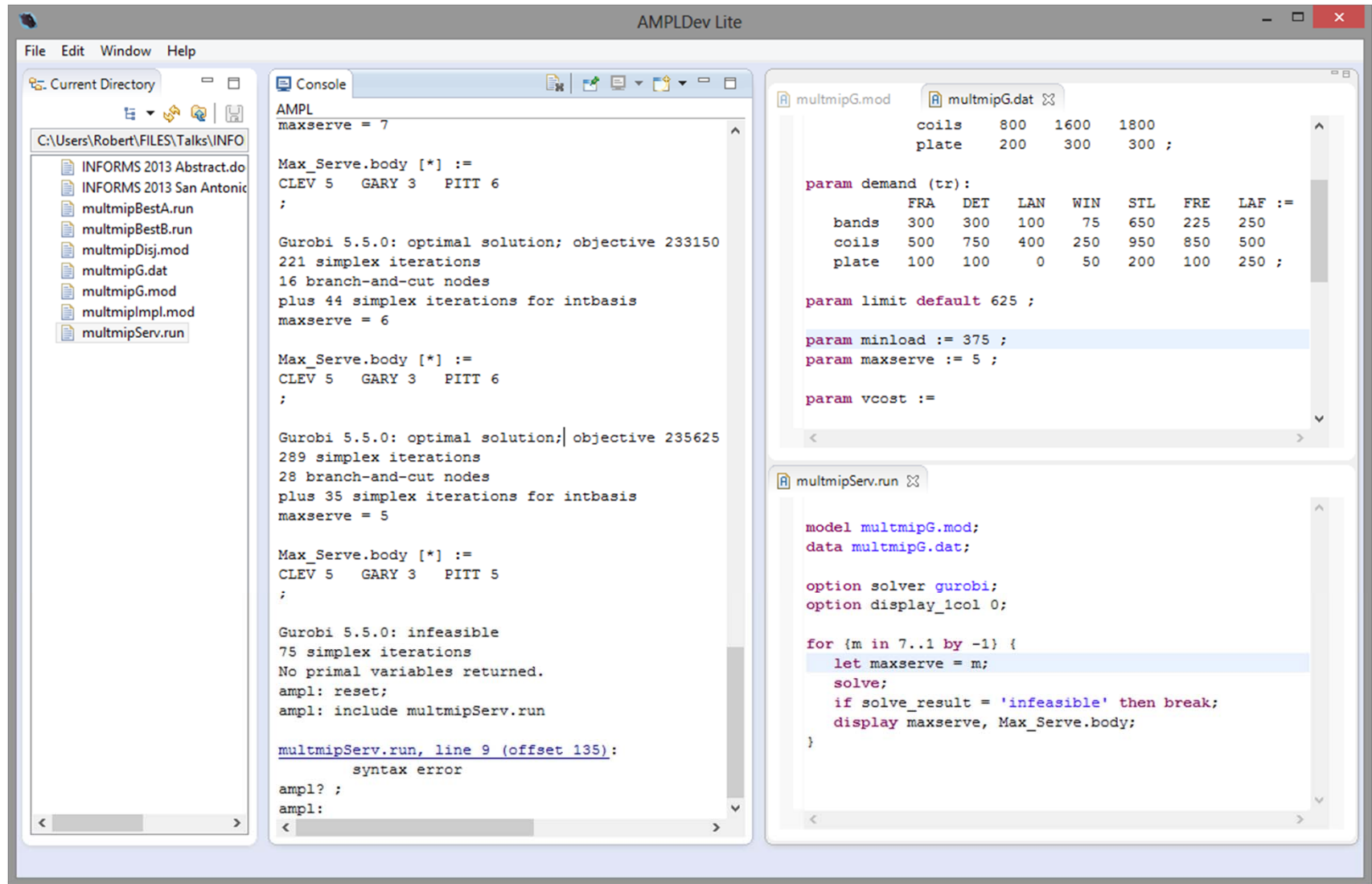
Integrated Development Environment

- ❖ Unified editor & command processor
- ❖ Included in the AMPL distribution
 - * Easy upgrade path
 - * Command-line, batch versions remain available
- ❖ Built on Eclipse
 - * Runs under Windows, Linux, MacOS

Initial release

- ❖ Simplified for easy transition
- ❖ Works with existing installations

Sample Screenshot



AMPL IDE

Version 1.0

Now available

- ❖ Available with all AMPL distributions
- ❖ Download add-on at www.ampl.com/IDE
 - * Integrated packages in preparation
- ❖ Version 1.01 updated with fixes available soon

Development details

- ❖ Partnership with OptiRisk Systems
- ❖ Long-term development & maintenance by AMPL
- ❖ “AMPLDEV” advanced IDE to be marketed by OptiRisk
 - * Offers full stochastic programming support

AMPL IDE

Version *x.y*

More help

- ❖ Option selection dialogs
 - * AMPL options
 - * Solver options
- ❖ AMPL language quick reference

NEOS Server access

Enhanced displays

- ❖ Parameter view windows
- ❖ Graphs

Suggestions from users . . .

www.ampl.com/newsite

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| | | | |
|--------|--------|--------|-------|
| CPLEX | Gurobi | Xpress | |
| CONOPT | KNITRO | MINOS | SNOPT |

AMPL
The AMPL system is a sophisticated modeling tool that supports the entire optimization modeling lifecycle: development, testing, deployment, and maintenance.
By using a high-level representation that represents optimization models in the same ways that people think about them, AMPL promotes rapid development and reliable results.
AMPL integrates a modeling language for describing optimization data, variables, objectives, and constraints; a command language for browsing models and analyzing results; and a scripting language for gathering and manipulating data and for implementing iterative

WHAT'S NEW?
March 6-8
6th INFORMS Optimization Society Conference, Houston, Texas.
AMPL sponsorship of this event & table in the exhibit area

FREE SOLVER TRIALS!

Readings (*AMPL*)

- ❖ R. Fourer, “Modeling Languages versus Matrix Generators for Linear Programming.” *ACM Transactions on Mathematical Software* **9** (1983) 143–183.
- ❖ R. Fourer, D.M. Gay, B.W. Kernighan, “A Modeling Language for Mathematical Programming.” *Management Science* **36** (1990) 519–554.
- ❖ Robert Fourer, “Database Structures for Mathematical Programming Models.” *Decision Support Systems* **20** (1997) 317–344.
- ❖ R. Fourer, D.M. Gay, B.W. Kernighan, *AMPL: A Modeling Language for Mathematical Programming*. Duxbury Press, Belmont, CA (first edition 1993, second edition 2003).
- ❖ Robert Fourer, On the Evolution of Optimization Modeling Systems. M. Groetschel (ed.), *Optimization Stories*. Documenta Mathematica (2012) 377-388.