



# Adding Optimization to Your Applications

Efficient and Effective Strategies  
Using Algebraic Modeling with AMPL

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# **Adding Optimization to Your Applications:**

## **Efficient and Effective Strategies Using Algebraic Modeling with AMPL**

Algebraic modeling languages have been developed with the goal of making optimization models much easier to develop, debug, and maintain. Moreover it is not necessary to give up these advantages when embedding a model into a larger system or deploying it to users. This talk describes features of the AMPL modeling language that work together to help you build optimization models quickly and use completed models effectively. Following an introduction to development of algebraic optimization models, the presentation introduces two distinct modeling language facilities commonly used to integrate models into projects and applications.:

- ❖ Scripting brings the programmer to the modeling language, extending the language so that the same constructs convenient for describing a model can also be used to specify how the model will be used in a broader context.
- ❖ APIs bring the modeling language to the programmer, providing access to model objects and methods for applications written in general-purpose programming languages.

The strengths of these approaches are contrasted using a common example.

# The Optimization Modeling Cycle

## *Steps*

- ❖ Communicate with problem owner
- ❖ Build model
- ❖ Prepare data
- ❖ Generate optimization problem
- ❖ Submit problem to solver
  - \* CPLEX, Gurobi, KNITRO, CONOPT, MINOS, . . .
- ❖ Report & analyze results
- ❖ ***Repeat!***

## *Goals for optimization software*

- ❖ Do this quickly and reliably
- ❖ Get results before client loses interest
- ❖ ***Deploy for application***

# Optimization Modeling Languages

## *Two forms of an optimization problem*

- ❖ Modeler's form
  - \* Mathematical description, easy for people to work with
- ❖ Algorithm's form
  - \* Explicit data structure, easy for solvers to compute with

## *Idea of a modeling language*

- ❖ ***A computer-readable modeler's form***
  - \* You write optimization problems in a modeling language
  - \* Computers translate to algorithm's form for solution

## *Advantages of a modeling language*

- ❖ Faster modeling cycles
- ❖ More reliable modeling
- ❖ More maintainable applications

# Algebraic Modeling Languages

## *Formulation concept*

- ❖ Define data in terms of sets & parameters
  - \* Analogous to database keys & records
- ❖ Define decision variables
- ❖ Minimize or maximize a function of decision variables
- ❖ Subject to equations or inequalities that constrain the values of the variables

## *Advantages*

- ❖ Familiar
- ❖ Powerful
- ❖ Implemented

# The AMPL Modeling Language

## *Features*

- ❖ Algebraic modeling language
- ❖ Variety of data sources
- ❖ Connections to all solver features

## *Advantages*

- ❖ Powerful, general expressions
- ❖ Natural, easy-to-learn design
- ❖ Efficient processing scales well with problem size

## ***3 ways to use***

- ❖ Command language
- ❖ Scripting language
- ❖ Programming interface (API)

# Outline

## *Simple roll cutting example*

- ❖ Solution via command language
- ❖ Sensitivity analysis via scripting

## *Roll cutting by pattern enumeration*

- ❖ via scripting
- ❖ via MATLAB API
- ❖ via Java API

## *Roll cutting by pattern generation*

- ❖ via scripting
- ❖ via MATLAB API

# Roll Cutting Problem

## *Motivation*

- ❖ Fill orders for rolls of various widths
  - \* by cutting raw rolls of one (large) fixed width
  - \* using a variety of cutting patterns

## *Optimization model*

- ❖ Decision variables
  - \* number of raw rolls to cut according to each pattern
- ❖ Objective
  - \* minimize number of raw rolls used
- ❖ Constraints
  - \* meet demands for each ordered width



*Roll cutting*

# Mathematical Formulation

*Given*

$W$  set of ordered widths

$n$  number of patterns considered

*and*

$a_{ij}$  occurrences of width  $i$  in pattern  $j$ ,  
for each  $i \in W$  and  $j = 1, \dots, n$

$b_i$  orders for width  $i$ , for each  $i \in W$

*Roll cutting*

## **Mathematical Formulation** (*cont'd*)

*Determine*

$X_j$  number of rolls to cut using pattern  $j$ ,  
for each  $j = 1, \dots, n$

*to minimize*

$$\sum_{j=1}^n X_j$$

total number of rolls cut

*subject to*

$$\sum_{j=1}^n a_{ij} X_j \geq b_i, \text{ for all } i \in W$$

number of rolls of width  $i$  cut  
must be at least the number ordered

*Roll Cutting*

# AMPL Formulation

*Symbolic model*

```
set WIDTHS;  
param orders {WIDTHS} > 0;  
param nPAT integer >= 0;  
param nbr {WIDTHS,1..nPAT} integer >= 0;  
  
var Cut {1..nPAT} integer >= 0;  
  
minimize Number:  
    sum {j in 1..nPAT} Cut[j];  
  
subj to Fulfill {i in WIDTHS}:  
    sum {j in 1..nPAT} nbr[i,j] * Cut[j] >= orders[i];
```

$$\sum_{j=1}^n a_{ij} X_j \geq b_i$$

*Roll Cutting*

# AMPL Formulation (*cont'd*)

## *Explicit data (independent of model)*

```
param: WIDTHS: orders :=
    6.77    10
    7.56    40
    17.46   33
    18.76   10 ;

param nPAT := 9 ;

param nbr:  1  2  3  4  5  6  7  8  9 :=
    6.77    0  1  1  0  3  2  0  1  4
    7.56    1  0  2  1  1  4  6  5  2
    17.46   0  1  0  2  1  0  1  1  1
    18.76   3  2  2  1  1  1  0  0  0 ;
```

# Command Language

*Model + data = problem instance to be solved*

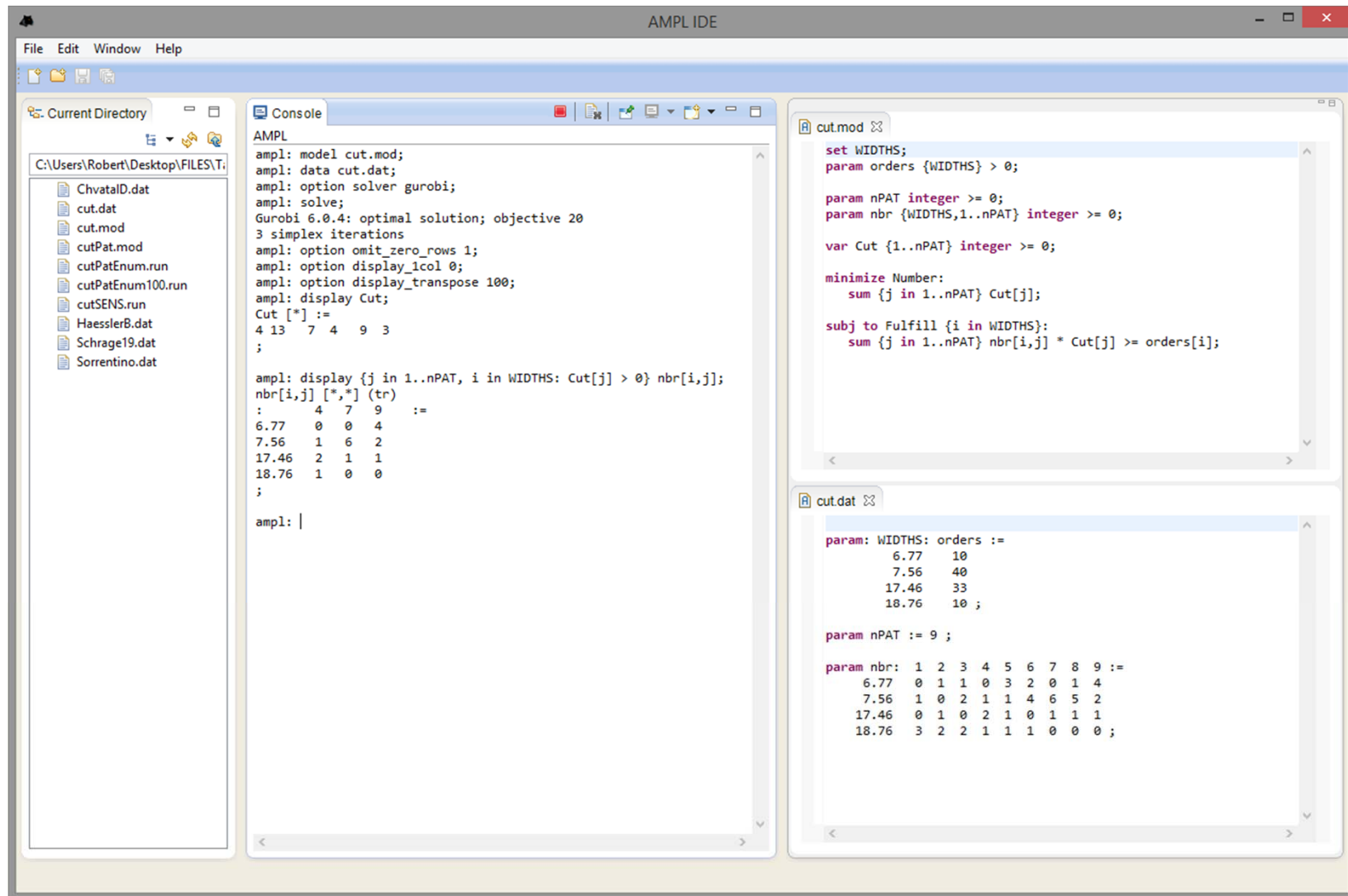
```
ampl: model cut.mod;
ampl: data cut.dat;
ampl: option solver gurobi;
ampl: solve;
Gurobi 6.0.4: optimal solution; objective 20
3 simplex iterations
ampl: display Cut;
4 13    7 4    9 3
ampl: display {j in 1..nPAT, i in WIDTHS: Cut[j] > 0} nbr[i,j];
:      4    7    9    :=
6.77   0    0    4
7.56   1    6    2
17.46  2    1    1
18.76  1    0    0
```

# Command Language (*cont'd*)

*Solver choice independent of model and data*

```
ampl: model cut.mod;
ampl: data cut.dat;
ampl: option solver cplex;
ampl: solve;
CPLEX 12.6.2.0: optimal integer solution; objective 20
3 MIP simplex iterations
ampl: display Cut;
4 13   7 4   9 3
ampl: display {j in 1..nPAT, i in WIDTHS: Cut[j] > 0} nbr[i,j];
:      4   7   9   :=
6.77   0   0   4
7.56   1   6   2
17.46  2   1   1
18.76  1   0   0
```

# IDE for Command Language



The screenshot displays the AMPL IDE interface. On the left is a file explorer showing the current directory: C:\Users\Robert\Desktop\FILES\T. The central console window shows the execution of an AMPL model, including the command 'ampl: solve;' and the output 'Gurobi 6.0.4: optimal solution; objective 20'. The console also displays the results of 'ampl: display Cut;' and 'ampl: display {j in 1..nPAT, i in WIDTHS: Cut[j] > 0} nbr[i,j];'. On the right, two code editors are visible. The top editor, 'cut.mod', contains the AMPL model code, including parameter declarations for WIDTHS, nPAT, and nbr, and a minimize objective function. The bottom editor, 'cut.dat', contains the data for the model, including the values for WIDTHS, nPAT, and the matrix nbr.

```
File Edit Window Help
Current Directory
C:\Users\Robert\Desktop\FILES\T
ChvatalD.dat
cut.dat
cut.mod
cutPat.mod
cutPatEnum.run
cutPatEnum100.run
cutSENS.run
HaesslerB.dat
Schrage19.dat
Sorrentino.dat

Console
AMPL
ampl: model cut.mod;
ampl: data cut.dat;
ampl: option solver gurobi;
ampl: solve;
Gurobi 6.0.4: optimal solution; objective 20
3 simplex iterations
ampl: option omit_zero_rows 1;
ampl: option display_1col 0;
ampl: option display_transpose 100;
ampl: display Cut;
Cut [*] :=
4 13 7 4 9 3
;

ampl: display {j in 1..nPAT, i in WIDTHS: Cut[j] > 0} nbr[i,j];
nbr[i,j] [*,*] (tr) :=
:      4 7 9
6.77  0 0 4
7.56  1 6 2
17.46 2 1 1
18.76 1 0 0
;

ampl: |

cut.mod
set WIDTHS;
param orders {WIDTHS} > 0;

param nPAT integer >= 0;
param nbr {WIDTHS,1..nPAT} integer >= 0;

var Cut {1..nPAT} integer >= 0;

minimize Number:
    sum {j in 1..nPAT} Cut[j];

subj to Fulfill {i in WIDTHS}:
    sum {j in 1..nPAT} nbr[i,j] * Cut[j] >= orders[i];

cut.dat
param: WIDTHS: orders :=
    6.77  10
    7.56  40
    17.46 33
    18.76  10 ;

param nPAT := 9 ;

param nbr:  1 2 3 4 5 6 7 8 9 :=
    6.77  0 1 1 0 3 2 0 1 4
    7.56  1 0 2 1 1 4 6 5 2
    17.46 0 1 0 2 1 0 1 1 1
    18.76 3 2 2 1 1 1 0 0 0 ;
```

# Scripting

*Bring the programmer to the modeling language*

*Extend modeling language syntax . . .*

- ❖ Algebraic expressions
- ❖ Set indexing expressions
- ❖ Interactive commands

*. . . with programming concepts*

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



*Scripting*

## **Parametric Analysis**

*Increase order levels in 1% steps*

- ❖ Increase orders[i] for all i in WIDTHS
- ❖ Record results
  - \* increments at which objective value increases
  - \* corresponding numbers of rolls cut

*. . . display results at the end*

*Scripting*

## Parametric Analysis (*cont'd*)

### *Script (setup)*

```
model cut.mod;
data cut.dat;

param prevNumber default
    (sum {i in WIDTHS} i * orders[i]) /
    (max {j in 1..nPAT} sum {i in WIDTHS} i * nbr[i,j]);

param baseOrders {WIDTHS};
let {i in WIDTHS} baseOrders[i] := orders[i];

set INCR default {}; # increments at which the objective changes
param incrObj {INCR}; # corresponding objective values

option solver Gurobi;
option solver_msg 0;
```

*Scripting*

## Parametric Analysis (*cont'd*)

*Script (looping and reporting)*

```
for {frac in 1 .. 1.25 by 0.01} {
  let {i in WIDTHS} orders[i] := frac * baseOrders[i];
  solve >Nul;

  if Number > prevNumber then {
    let INCR := INCR union {frac};
    let incrObj[frac] := Number;
    let prevNumber := Number;
  }
}

printf ' Step  Number\n';
printf {frac in INCR}: '%5.2f%5d\n', frac, incrObj[frac];
```

*Scripting*

# Parametric Analysis (*cont'd*)

*Script run*

```
AMPL: include cutSens.run;
```

Step	Number
------	--------

1.00	20
------	----

1.01	21
------	----

1.07	22
------	----

1.13	23
------	----

1.16	24
------	----

1.22	25
------	----

```
AMPL:
```

*Scripting*

## **Cutting *via* Pattern Enumeration**

*Build the pattern list, then solve*

- ❖ Read general model
- ❖ Read data: demands, raw width
- ❖ Compute data: all usable patterns
- ❖ Solve problem instance

*Scripting*

# Pattern Enumeration

*Model*

```
param roll_width > 0;
set WIDTHS ordered by reversed Reals;
param orders {WIDTHS} > 0;

param maxPAT integer >= 0;
param nPAT integer >= 0, <= maxPAT;

param nbr {WIDTHS,1..maxPAT} integer >= 0;

var Cut {1..nPAT} integer >= 0;

minimize Number:
    sum {j in 1..nPAT} Cut[j];

subj to Fulfill {i in WIDTHS}:
    sum {j in 1..nPAT} nbr[i,j] * Cut[j] >= orders[i];
```

*Scripting*

# Pattern Enumeration

*Data*

```
param roll_width := 64.50 ;  
param: WIDTHS: orders :=  
    6.77    10  
    7.56    40  
    17.46   33  
    18.76   10 ;
```

*Scripting*

# Pattern Enumeration

*Script (initialize)*

```
model cutPAT.mod;
data Sorrentino.dat;

param curr_sum >= 0;
param curr_width > 0;
param pattern {WIDTHS} integer >= 0;

let maxPAT := 1000000;

let nPAT := 0;
let curr_sum := 0;
let curr_width := first(WIDTHS);
let {w in WIDTHS} pattern[w] := 0;
```



*Scripting*

# Pattern Enumeration

*Script (loop)*

```
repeat {
  if curr_sum + curr_width <= roll_width then {
    let pattern[curr_width] := floor((roll_width-curr_sum)/curr_width);
    let curr_sum := curr_sum + pattern[curr_width] * curr_width;
  }
  if curr_width != last(WIDTHS) then
    let curr_width := next(curr_width,WIDTHS);
  else {
    let nPAT := nPAT + 1;
    let {w in WIDTHS} nbr[w,nPAT] := pattern[w];
    let curr_sum := curr_sum - pattern[last(WIDTHS)] * last(WIDTHS);
    let pattern[last(WIDTHS)] := 0;
    let curr_width := min {w in WIDTHS: pattern[w] > 0} w;
    if curr_width < Infinity then {
      let curr_sum := curr_sum - curr_width;
      let pattern[curr_width] := pattern[curr_width] - 1;
      let curr_width := next(curr_width,WIDTHS);
    }
    else break;
  }
}
```

*Scripting*

# Pattern Enumeration

*Script (solve, report)*

```
option solver gurobi;
solve;
printf "\n%5i patterns, %3i rolls", nPAT, sum {j in 1..nPAT} Cut[j];
printf "\n\n Cut  ";
printf {j in 1..nPAT: Cut[j] > 0}: "%3i", Cut[j];
printf "\n\n";
for {i in WIDTHS} {
    printf "%7.2f ", i;
    printf {j in 1..nPAT: Cut[j] > 0}: "%3i", nbr[i,j];
    printf "\n";
}
printf "\nWASTE = %5.2f%\n\n",
    100 * (1 - (sum {i in WIDTHS} i * orders[i]) / (roll_width * Number));
```

*Scripting*

# Pattern Enumeration

## *Results*

```
ampl: include cutPatEnum.run
```

```
Gurobi 5.6.0: optimal solution; objective 18
```

```
7 simplex iterations
```

**43 patterns, 18 rolls**

```
Cut      2  2  3 11
```

```
18.76   3  2  0  0
```

```
17.46   0  1  3  2
```

```
7.56    1  1  1  3
```

```
6.77    0  0  0  1
```

```
WASTE = 2.34%
```

*Scripting*

# Pattern Enumeration

## *Data 2*

```
param roll_width := 349 ;  
param: WIDTHS: orders :=  
    28.75    7  
    33.75    23  
    34.75    23  
    37.75    31  
    38.75    10  
    39.75    39  
    40.75    58  
    41.75    47  
    42.25    19  
    44.75    13  
    45.75    26 ;
```

*Scripting*

# Pattern Enumeration

## *Results 2*

```
ampl: include cutPatEnum.run
```

```
Gurobi 4.6.1: optimal solution; objective 34
```

```
291 simplex iterations
```

```
54508 patterns, 34 rolls
```

Cut	8	1	1	1	3	1	1	1	1	2	7	2	3	1	1
45.75	3	2	0	0	0	0	0	0	0	0	0	0	0	0	0
44.75	1	2	2	1	0	0	0	0	0	0	0	0	0	0	0
42.25	0	2	0	0	4	2	2	1	0	0	0	0	0	0	0
41.75	4	2	0	2	0	0	0	0	2	1	1	0	0	0	0
40.75	0	0	4	4	1	4	3	0	2	3	1	6	3	2	2
39.75	0	0	0	0	0	0	0	2	0	0	5	0	0	2	0
38.75	0	0	1	0	0	0	0	0	4	0	0	0	0	2	3
37.75	0	0	0	0	0	0	1	0	0	4	0	0	6	2	4
34.75	0	0	0	0	4	0	3	1	0	0	0	3	0	1	0
33.75	0	0	0	0	0	3	0	4	0	1	2	0	0	0	0
28.75	0	0	2	2	0	0	0	2	1	0	0	0	0	0	0

```
WASTE = 0.69%
```

*Scripting*

# Pattern Enumeration

## *Data 3*

```
param roll_width := 172 ;  
param: WIDTHS: orders :=  
    25.000    5  
    24.750    73  
    18.000    14  
    17.500    4  
    15.500    23  
    15.375    5  
    13.875    29  
    12.500    87  
    12.250    9  
    12.000    31  
    10.250    6  
    10.125    14  
    10.000    43  
    8.750     15  
    8.500     21  
    7.750     5 ;
```

*Scripting*

# Pattern Enumeration

*Results 3 (using a subset of patterns)*

```
ampl: include cutPatEnum.run
```

```
Gurobi 4.6.1: optimal solution; objective 33
```

```
722 simplex iterations
```

```
40 branch-and-cut nodes
```

```
273380 patterns, 33 rolls
```

Cut	1	1	1	1	4	4	4	1	1	2	5	2	1	1	1	3
25.00	2	1	1	1	0	0	0	0	0	0	0	0	0	0	0	0
24.75	1	2	1	0	5	4	3	2	2	2	2	1	1	0	0	0
18.00	0	0	0	0	1	0	0	1	0	0	0	1	1	5	1	0
17.50	0	3	0	0	0	0	0	0	0	0	0	0	0	0	1	0
.....																
10.12	0	2	0	0	0	1	2	0	0	0	0	0	0	0	0	0
10.00	0	0	0	0	0	2	0	1	3	0	6	0	0	2	0	0
8.75	0	0	1	0	0	0	0	0	0	2	0	2	0	0	0	2
8.50	0	0	2	0	0	2	0	0	0	0	0	4	3	0	0	0
7.75	0	0	0	0	1	0	0	1	0	0	0	0	0	0	0	0

```
WASTE = 0.62%
```

*Scripting*

## **Cutting via Pattern Generation**

*Generate the pattern list by a series of solves*

- ❖ Solve LP relaxation using subset of patterns
- ❖ Add “most promising” pattern to the subset
  - \* Minimize reduced cost given dual values
  - \* Equivalent to a knapsack problem
- ❖ Iterate as long as there are promising patterns
  - \* Stop when minimum reduced cost is zero
- ❖ Solve IP using all patterns found



*Scripting*

# Pattern Generation

## *Cutting model*

```
set WIDTHS ordered by reversed Reals;
param orders {WIDTHS} > 0;

param nPAT integer >= 0, <= maxPAT;
param nbr {WIDTHS,1..nPAT} integer >= 0;

var Cut {1..nPAT} integer >= 0;

minimize Number:
    sum {j in 1..nPAT} Cut[j];

subj to Fulfill {i in WIDTHS}:
    sum {j in 1..nPAT} nbr[i,j] * Cut[j] >= orders[i];
```

*Scripting*

# Pattern Generation

## *Knapsack model*

```
param roll_width > 0;
param price {WIDTHS} default 0.0;

var Use {WIDTHS} integer >= 0;

minimize Reduced_Cost:
    1 - sum {i in WIDTHS} price[i] * Use[i];

subj to Width_Limit:
    sum {i in WIDTHS} i * Use[i] <= roll_width;
```

*Scripting*

# Pattern Generation

*Script (problems, initial patterns)*

```
model cutPatGen.mod;
data Sorrentino.dat;

problem Cutting_Opt: Cut, Number, Fill;
    option relax_integrality 1;
    option presolve 0;

problem Pattern_Gen: Use, Reduced_Cost, Width_Limit;
    option relax_integrality 0;
    option presolve 1;

let nPAT := 0;
for {i in WIDTHS} {
    let nPAT := nPAT + 1;
    let nbr[i,nPAT] := floor (roll_width/i);
    let {i2 in WIDTHS: i2 <> i} nbr[i2,nPAT] := 0;
};
```

*Scripting*

# Pattern Generation

*Script (generation loop)*

```
repeat {
  solve Cutting_Opt;
  let {i in WIDTHS} price[i] := Fill[i].dual;
  solve Pattern_Gen;
  printf "\n%7.2f%11.2e  ", Number, Reduced_Cost;
  if Reduced_Cost < -0.00001 then {
    let nPAT := nPAT + 1;
    let {i in WIDTHS} nbr[i,nPAT] := Use[i];
  }
  else break;
  for {i in WIDTHS} printf "%3i", Use[i];
};
```

*Scripting*

# Pattern Generation

*Script (final integer solution)*

```
option Cutting_Opt.relax_integrality 0;
option Cutting_Opt.presolve 10;
solve Cutting_Opt;

if Cutting_Opt.result = "infeasible" then
  printf "\n*** No feasible integer solution ***\n\n";
else {
  printf "Best integer: %3i rolls\n\n", sum {j in 1..nPAT} Cut[j];
  for {j in 1..nPAT: Cut[j] > 0} {
    printf "%3i of:", Cut[j];
    printf {i in WIDTHS: nbr[i,j] > 0}: "%3i x %6.3f", nbr[i,j], i;
    printf "\n";
  }

  printf "\nWASTE = %5.2f%%\n\n",
    100 * (1 - (sum {i in WIDTHS} i * orders[i]) / (roll_width * Number));
}
```

*Scripting*

# Pattern Generation

*Results (relaxation)*

```
ampl: include cutpatgen.run
```

```
20.44 -1.53e-01 1 3 2 0
18.78 -1.11e-01 0 1 3 0
18.37 -1.25e-01 0 1 0 3
17.96 -4.17e-02 0 6 0 1
17.94 -1.00e-06
```

Optimal relaxation: **17.9412 rolls**

```
10.0000 of: 1 x 6.770 3 x 7.560 2 x 17.460
4.3333 of: 1 x 7.560 3 x 17.460
3.1961 of: 1 x 7.560 3 x 18.760
0.4118 of: 6 x 7.560 1 x 18.760
```

WASTE = 2.02%

*Scripting*

# Pattern Generation

*Results (integer)*

Rounded up to integer: **20 rolls**

Cut	10	5	4	1
6.77	1	0	0	0
7.56	3	1	1	6
17.46	2	3	0	0
18.76	0	0	3	1

WASTE = 12.10%

Best integer: **19 rolls**

Cut	10	5	3	1
6.77	1	0	0	0
7.56	3	1	1	6
17.46	2	3	0	0
18.76	0	0	3	1

WASTE = 7.48%

*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
- ❖ Complex set & data structures

### *Expressiveness*

- ❖ Based on a declarative language
- ❖ Not object-oriented

*So . . .*

# **APIs** (application programming interfaces)

*Bring the modeling language to the programmer*

- ❖ Data and result management in a general-purpose programming language
- ❖ Modeling and solving through calls to AMPL

*Development details*

- ❖ Partnership with OptiRisk Systems
  - \* Christian Valente, principal developer
- ❖ Long-term development & maintenance by AMPL
  - \* Victor Zverovich, project coordinator

# Cutting Revisited

## *Hybrid approach*

- ❖ Control & pattern creation from a programming language
  - \* Pattern enumeration: finding all patterns
  - \* Pattern generation: solving knapsack problems
- ❖ Model & modeling commands in AMPL

## *Two programming languages*

- ❖ Java
- ❖ MATLAB

## *Key to examples*

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

# AMPL Model File

## *Basic pattern-cutting model*

```
param nPatterns integer > 0;

set PATTERNS = 1..nPatterns; # patterns
set WIDTHS; # finished widths

param order {WIDTHS} >= 0; # rolls of width j ordered
param overrun; # permitted overrun on any width

param rolls {WIDTHS,PATTERNS} >= 0; # rolls of width i in pattern j

var Cut {PATTERNS} integer >= 0; # raw rolls to cut in each pattern

minimize TotalRawRolls: sum {p in PATTERNS} Cut[p];

subject to FinishedRollLimits {w in WIDTHS}:
    order[w] <= sum {p in PATTERNS} rolls[w,p] * Cut[p] <= order[w] + overrun;
```

# Pattern Enumeration in MATLAB

*Load & generate data, set up AMPL model*

```
function cuttingEnum(dataFile)

% Get data from .mat file: roll_width, overrun, widths, orders
load(dataFile);

% Generate pattern matrix
[widthsDec,ind] = sort(widths,'descend');
patmat = patternEnum(roll_width,widthsDec);
patmat(:,ind) = patmat;

% Initialize and load cutting-stock model from file
AMPL = AMPL();
AMPL.read('cut.mod');
```

# Pattern Enumeration in MATLAB

## *Send data to AMPL*

```
% Send scalar values
AMPL.getParameter('overrun').setValues(overrun);
AMPL.getParameter('nPatterns').setValues(length(patmat));

% Send order vector
WidthOrder = DataFrame(1, 'WIDTHS', 'order');
WidthOrder.setColumn('WIDTHS', num2cell(widths));
WidthOrder.setColumn('order', orders);
AMPL.setData(WidthOrder, 'WIDTHS');

% Send pattern matrix
AllPatterns = DataFrame(2, 'WIDTHS', 'PATTERNS', 'rolls');
AllPatterns.setMatrix('patmat', num2cell(widths), num2cell(1:length(patmat)));
AMPL.setData(AllPatterns)
```

# Pattern Enumeration in MATLAB

## *Solve and report*

```
% Solve
ampl.setOption('solver' , 'gurobi');
ampl.solve

% Retrieve solution
CuttingPlan = ampl.getVariable('Cut').getValues();
cutvec = CuttingPlan.getColumnAsDoubles('val');

% Display solution
cuttingPlot (roll_width, widths, patmat(cutvec>0,:), cutvec(cutvec>0))
```

# Pattern Enumeration in MATLAB

## *Enumeration routine*

```
function patmat = patternEnum(rollwidth,widths)
if length(widths) == 1
    patmat = floor(rollwidth/widths(1));
else
    patmat = [];
    for n = floor(rollwidth/widths(1)):-1:0
        patnew = patternEnum (rollwidth-n*widths(1), widths(2:end));
        patmat = [patmat; n*ones(size(patnew,1),1) patnew];
    end
end
```



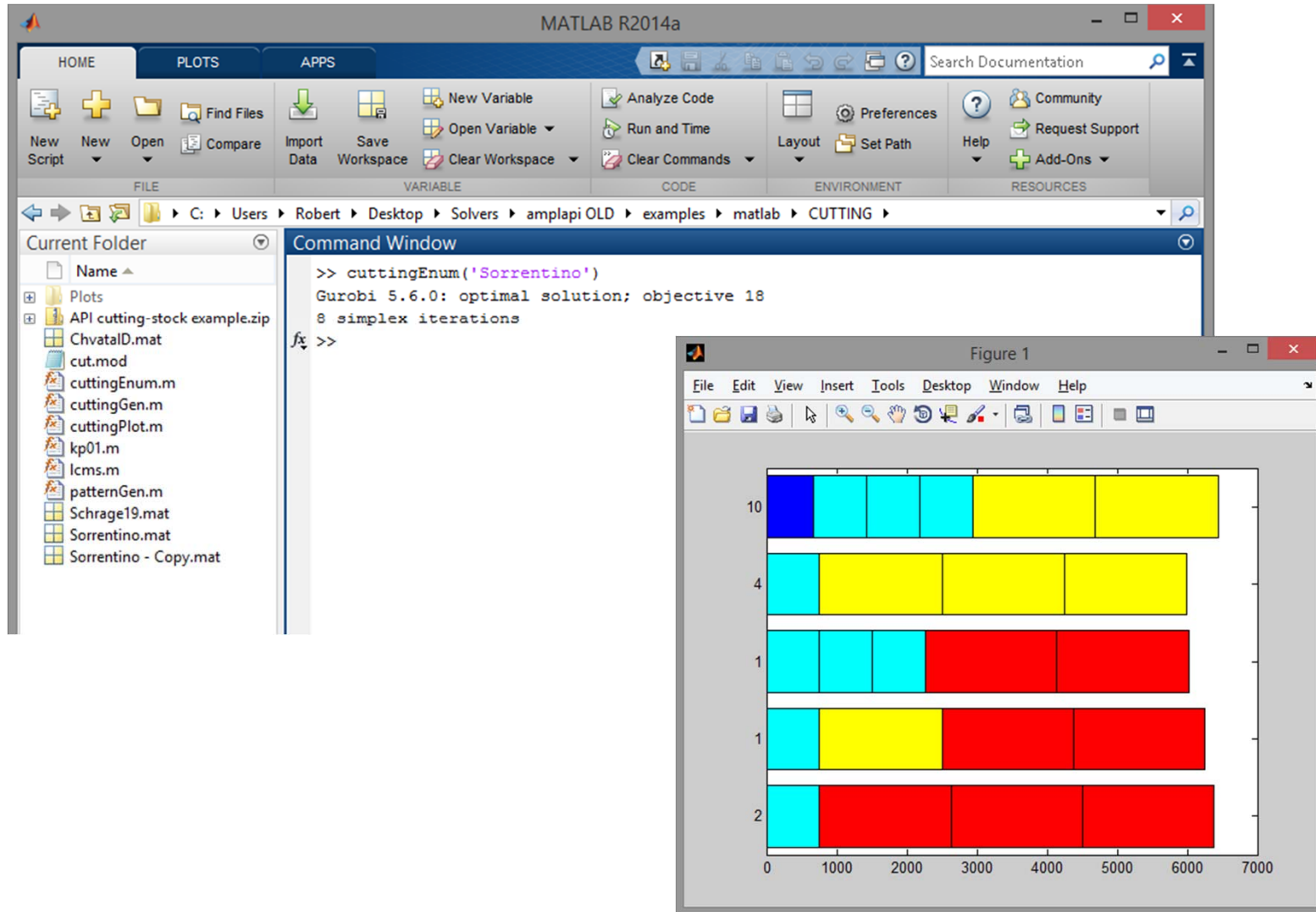
# Pattern Enumeration in MATLAB

## *Plotting routine*

```
function cuttingPlot (roll_width,widths,patmat,cutvec)
plotmat = zeros(length(cutvec),sum(max(patmat)));
colors = jet(length(widths));
plotpos = 0;
for j = 1:length(widths)
    for i = 1:length(cutvec)
        plotmat(i,plotpos+1:plotpos+patmat(i,j)) = widths(j);
    end
    for i = 1:max(patmat(:,j))
        colormat(plotpos+i,:) = colors(j,:);
    end
    plotpos = plotpos + max(patmat(:,j));
end
colormap(colormat); shading faceted
h = barh(plotmat,'stacked');
set (h, 'edgecolor','black')
set(gca, 'YTickLabel',num2cell(cutvec))
```

AMPL API

# Pattern Enumeration in MATLAB



# Pattern Enumeration in Java

*Generate patterns, set up AMPL model*

```
public static void main(String[] args) throws IOException {
    import static com.ampl.examples.CuttingStock.Sorrentino;
    int[] sortedWidths = widths.clone();
    sortDescending(sortedWidths);
    ArrayList<Integer> patterns = new ArrayList<>();
    patternEnum (roll_width, sortedWidths, 0, patterns);

    // Initialize and load cutting-stock model from file
    AMPL ampl = new AMPL();
    try {
        ampl.read("cut.mod");
    }
}
```

# Pattern Enumeration in Java

## *Send data to AMPL*

```
ampl.getParameter("overrun").setValues(overrun);
int numPatterns = patterns.size() / widths.length;
ampl.getParameter("nPatterns").setValues(numPatterns);

DataFrame widthOrder = new DataFrame(1, "WIDTHS", "order");
widthOrder.setColumn("WIDTHS", widths);
widthOrder.setColumn("order", orders);
ampl.setData(widthOrder, true);

DataFrame allPatterns = new DataFrame(2, "WIDTHS", "PATTERNS", "rolls");
for (int i = 0; i < widths.length; i++) {
    for (int j = 0; j < numPatterns; j++) {
        allPatterns.addRow(
            sortedWidths[i], j + 1, patterns.get(j * widths.length + i));
    }
}
ampl.setData(allPatterns, false);
```

*AMPL API*

# Pattern Enumeration in Java

*Solve and report solution*

```
    ampl.setOption("solver", "gurobi");
    ampl.solve();
    printSolution (ampl.getVariable("Cut"), ampl.getParameter("rolls"));
} finally {
    ampl.close();
}
}
```

*AMPL API*

# Pattern Generation in MATLAB

*Set up AMPL, get data*

```
function cuttingGen(dataFile)

% Initialize
AMPL = AMPL();

% Load cutting-stock model from file
AMPL.read('cut.mod');
Cut = AMPL.getVariable('Cut');
Limits = AMPL.getConstraint('FinishedRollLimits');

% Get data from .mat file: roll_width, overrun, widths, orders
load(dataFile);
```

# Pattern Generation in MATLAB

## *Send data to AMPL*

```
% Send scalar values
```

```
AMPL.getParameter('overrun').setValues(overrun);  
AMPL.getParameter('nPatterns').setValues(length(widths));
```

```
% Send order vector
```

```
WidthOrder = DataFrame(1, 'WIDTHS', 'order');  
WidthOrder.setColumn('WIDTHS', num2cell(widths));  
WidthOrder.setColumn('order', orders);  
AMPL.setData(WidthOrder, 'WIDTHS');
```

```
% Generate and send initial pattern matrix
```

```
maxpat = floor(roll_width./widths);  
patmat = diag(maxpat);  
InitPatterns = DataFrame(2, 'WIDTHS', 'PATTERNS', 'rolls');  
InitPatterns.setMatrix(patmat, num2cell(widths), num2cell(1:length(widths)));  
AMPL.setData(InitPatterns);
```

# Pattern Generation in MATLAB

## *Set up for generation loop*

```
% Set solve options
AMPL.setOption('solver','gurobi');
AMPL.setOption('relax_integrality','1');

% Set up DataFrame for sending AMPL new patterns
AMPL.eval('param newpat {WIDTHS} integer >= 0;');
NewPattern = DataFrame(1, 'WIDTHS', 'newpat');
NewPattern.setColumn('WIDTHS', num2cell(widths));

% Compute multiplier for integer weights
[n,d] = rat(widths);
intmult = lcm(d);
```



# Pattern Generation in MATLAB

*Loop 1: Retrieve duals & look for new pattern*

```
while 1
    ampl.solve
    DualPrices = Limits.getValues;
    dualvec = DualPrices.getColumnAsDoubles('dual');
    wgt = []; val = [];
    for w = 1:length(widths)
        if dualvec(w) > 0
            wgt = [wgt widths(w)*ones(1,maxpat(w))];
            val = [val dualvec(w)*ones(1,maxpat(w))];
        end
    end
    % Solve knapsack problem for potential new pattern
    [kmax,z] = kp01 (round(intmult*wgt), val, intmult*roll_width);
    if kmax < 1.000001
        break;
    end
end
```

# Pattern Generation in MATLAB

## *Loop 2: Send new pattern to AMPL*

```
widthlist = wgt(z);
for w = 1:length(widths)
    newpat(w) = length(find(widthlist==widths(w)));
end
patmat = [patmat; newpat];
NewPattern.setColumn('newpat', newpat);
ampl.setData(NewPattern);

ampl.eval('let nPatterns := nPatterns + 1;');
ampl.eval('let {w in WIDTHS} rolls[w,nPatterns] := newpat[w];');
end

% Compute and display integer solution
ampl.setOption('relax_integrality','0');
ampl.solve;

CuttingPlan = Cut.getValues();
cutvec = CuttingPlan.getColumnAsDoubles('val');
cuttingPlot (roll_width, widths, patmat(cutvec>0,:), cutvec(cutvec>0))
```

# Pattern Generation in MATLAB

The image shows the MATLAB R2014a environment. The Command Window displays the following output:

```
>> cuttingGen('Sorrentino')  
Gurobi 5.6.0: optimal solution; objective 20.44444444  
Gurobi 5.6.0: optimal solution; objective 18.79166667  
Gurobi 5.6.0: optimal solution; obj  
5 simplex iterations  
Gurobi 5.6.0: optimal solution; obj  
5 simplex iterations  
Gurobi 5.6.0: optimal solution; obj  
5 simplex iterations  
Gurobi 5.6.0: optimal solution; obj  
3 simplex iterations  
fx >>
```

The Figure window, titled 'Figure 1', shows a horizontal bar chart with the following data series:

Iteration Count	Color
1	Cyan
4	Yellow
10	Blue
4	Red

## **Data Transfer: Alternatives**

### *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

### *Scripting: 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*

*AMPL API*

## **Availability**

*Java API version 1.0 released*

*MATLAB API version 1.0 released*

- ❖ Add-ons to all AMPL distributions
- ❖ Download from [www.ampl.com/products/api/](http://www.ampl.com/products/api/)

*C++ API in final development*

- ❖ Release planned for this summer

*More languages to follow*

- ❖ R
- ❖ Python
- ❖ .NET: C#, Visual Basic

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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

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# AMPL Readings

- ❖ 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.