

# Alternatives for Scripting in the AMPL Modeling Language



*Robert Fourer*

AMPL Optimization

[www.ampl.com](http://www.ampl.com) — 773-336-AMPL

Industrial Engineering & Management Sciences,  
Northwestern University

**4<sup>th</sup> INFORMS Optimization Society Conference**

Coral Gables, Florida — 24-26 February, 2012

Session SB01, *Software for Optimization Modeling*

# Topics

**1:** *Parametric analysis*

**2:** *Solution generation*

**a:** *via cuts*

**b:** *via solver*

**3:** *Heuristic optimization*

**4:** *Pattern generation*

**5:** *Decomposition*

*Scripts in practice . . .*

*Prospective improvements . . .*

# 1: Parametric Analysis

## *Multi-period production*

- ❖ Max total profit
- ❖ Production, inventory, sales variables
- ❖ Limited production time per period

## *Try different time availabilities in period 3*

- ❖ Step availability and re-solve
  - \* until dual is zero (constraint is slack)
- ❖ Record results
  - \* distinct dual values
  - \* corresponding objective values

# Parametric

## *Model (sets, parameters, variables)*

```
set PROD;      # products
param T > 0;   # number of weeks

param rate {PROD} > 0;      # tons per hour produced
param inv0 {PROD} >= 0;    # initial inventory
param avail {1..T} >= 0;   # hours available in week
param market {PROD,1..T} >= 0; # limit on tons sold in week

param prodcost {PROD} >= 0; # cost per ton produced
param invcost {PROD} >= 0;  # carrying cost/ton of inventory
param revenue {PROD,1..T} >= 0; # revenue per ton sold

var Make {PROD,1..T} >= 0;  # tons produced
var Inv {PROD,0..T} >= 0;  # tons inventoried
var Sell {p in PROD, t in 1..T} >= 0, <= market[p,t];
                                # tons sold
```

# Parametric

## *Model (objective, constraints)*

```
maximize Total_Profit:
    sum {p in PROD, t in 1..T} (revenue[p,t]*Sell[p,t] -
        prodcost[p]*Make[p,t] - invcost[p]*Inv[p,t]);

subject to Time {t in 1..T}:
    sum {p in PROD} (1/rate[p]) * Make[p,t] <= avail[t];

subject to Init_Inv {p in PROD}:
    Inv[p,0] = inv0[p];

subject to Balance {p in PROD, t in 1..T}:
    Make[p,t] + Inv[p,t-1] = Sell[p,t] + Inv[p,t];
```

# Parametric

## *Data*

```
param T := 4;
set PROD := bands coils;

param avail := 1 40 2 40 3 32 4 40 ;

param rate := bands 200 coils 140 ;
param inv0 := bands 10 coils 0 ;

param prodcost := bands 10 coils 11 ;
param invcost := bands 2.5 coils 3 ;

param revenue: 1 2 3 4 :=
    bands 25 26 27 27
    coils 30 35 37 39 ;

param market: 1 2 3 4 :=
    bands 6000 6000 4000 6500
    coils 4000 2500 3500 4200 ;
```

# Parametric

## *Script*

```
set AVAIL default {};  
param avail_obj {AVAIL};  
param avail_dual {AVAIL};  
  
let avail[3] := 1;  
param avail_step = 1;  
param previous_dual default Infinity;  
repeat while previous_dual > 0 {  
  solve;  
  if time[3].dual < previous_dual then {  
    let AVAIL := AVAIL union {avail[3]};  
    let avail_obj[avail[3]] := total_profit;  
    let avail_dual[avail[3]] := time[3].dual;  
    let previous_dual := time[3].dual;  
  }  
  
  let avail[3] := avail[3] + avail_step;  
}
```

# Parametric

## *Results*

```
ampl: include steeltparam.run;
ampl: display avail_obj, avail_dual;

:  avail_obj  avail_dual  :=
1   404616    3620
23  484233    3500
26  494633    3400
45  559233    2980
68  626283     0
;
```



# Parametric: Observations

## *Parameters are true objects*

- ❖ Assign new value to param avail[3]
  - \* `let avail[3] := avail[3] + avail_step;`
- ❖ Problem instance changes accordingly

## *Sets are true data*

- ❖ Assign new value to set AVAIL
  - \* `let AVAIL := AVAIL union {avail[3]};`
- ❖ All indexed entities change accordingly

## 2a: Solution Generation *via Cuts*

### *Workforce planning*

- ❖ Cover demands for workers
  - \* Each “shift” requires a certain number of employees
  - \* Each employee works a certain “schedule” of shifts
- ❖ Satisfy scheduling rules
  - \* Only “valid” schedules from given list may be used
  - \* *Each schedule that is used at all must be worked by at least ?? employees*
- ❖ Minimize total workers needed
  - \* Which schedules should be used?
  - \* How many employees should work each schedule?

### *Generate alternative optimal solutions*

- ❖ Save & display each shift schedule

# Solutions *via Cuts*

## *Model (sets, parameters)*

```
set SHIFTS;                # shifts
param Nsched;             # number of schedules;
set SCHEDS = 1..Nsched;   # set of schedules

set SHIFT_LIST {SCHEDS} within SHIFTS;

param rate {SCHEDS} >= 0;    # pay rates
param required {SHIFTS} >= 0; # staffing requirements
param least_assign >= 0;    # min workers on any schedule used
```

# Solutions *via Cuts*

*Model (variables, objective, constraints)*

```
var Work {SCHEDS} >= 0 integer;
var Use  {SCHEDS} >= 0 binary;

minimize Total_Cost:
    sum {j in SCHEDS} rate[j] * Work[j];

subject to Shift_Needs {i in SHIFTS}:
    sum {j in SCHEDS: i in SHIFT_LIST[j]} Work[j] >= required[i];

subject to Least_Use1 {j in SCHEDS}:
    least_assign * Use[j] <= Work[j];

subject to Least_Use2 {j in SCHEDS}:
    Work[j] <= (max {i in SHIFT_LIST[j]} required[i]) * Use[j];
```

# Solutions *via Cuts*

## *Data*

```
set SHIFTS := Mon1 Tue1 Wed1 Thu1 Fri1 Sat1
            Mon2 Tue2 Wed2 Thu2 Fri2 Sat2
            Mon3 Tue3 Wed3 Thu3 Fri3 ;

param Nsched := 126 ;

set SHIFT_LIST[1] := Mon1 Tue1 Wed1 Thu1 Fri1 ;
set SHIFT_LIST[2] := Mon1 Tue1 Wed1 Thu1 Fri2 ;
set SHIFT_LIST[3] := Mon1 Tue1 Wed1 Thu1 Fri3 ;
set SHIFT_LIST[4] := Mon1 Tue1 Wed1 Thu1 Sat1 ;
set SHIFT_LIST[5] := Mon1 Tue1 Wed1 Thu1 Sat2 ; .....
```

  

```
param required := Mon1 100 Mon2 78 Mon3 52
                  Tue1 100 Tue2 78 Tue3 52
                  Wed1 100 Wed2 78 Wed3 52
                  Thu1 100 Thu2 78 Thu3 52
                  Fri1 100 Fri2 78 Fri3 52
                  Sat1 100 Sat2 78 ;
```

# Solutions *via Cuts*

## *Script*

```
param nSols default 0;
param maxSols = 20;

set USED {1..nSols} within SCHEDS;

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

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

# Solutions *via Cuts*

## *Results*

```
ampl: include scheds.run
```

```
Gurobi 4.0.1: optimal solution; objective 266
```

```
857 simplex iterations
```

```
29 branch-and-cut nodes
```

```
Work [*] :=
```

```
  1 21    21 36    52  7    89 29    94  7    109 16    124 36  
  3  7    37 29    71 13    91 16    95 13    116 36 ;
```

```
Gurobi 4.0.1: optimal solution; objective 266
```

```
1368 simplex iterations
```

```
59 branch-and-cut nodes
```

```
Work [*] :=
```

```
  1  9    17  9    38  7    59 21    75 36    94  7    114  8    124 35  
  4 20    33 27    56  7    71 27    86  8    107  9    116 36 ;
```

# Solutions *via Cuts*

## Results (continued)

Gurobi 4.0.1: optimal solution; **objective 266**

982 simplex iterations

57 branch-and-cut nodes

Work [\*] :=

2	28	16	8	38	18	75	34	86	8	108	8	115	16	121	36
7	18	28	10	70	18	85	18	97	18	109	10	116	18		;

Gurobi 4.0.1: optimal solution; **objective 266**

144 simplex iterations

Work [\*] :=

2	29	16	7	76	36	88	29	106	16	116	7	123	7		
7	36	70	28	85	7	97	7	109	29	121	21	126	7		;

Gurobi 4.0.1: optimal solution; **objective 266**

122 simplex iterations

Work [\*] :=

2	15	16	20	70	15	85	21	106	16	116	21	123	21		
7	36	53	14	76	36	97	21	109	15	121	8	126	7		;



# Solutions *via Cuts*: Observations

*Same expressions describe sets and indexing*

- ❖ Index a summation
  - \* ... `sum {j in SCHEDULES diff USED[k]} Use[j] >= 1;`
- ❖ Assign a value to a set
  - \* `let USED[nSols] := {j in SCHEDULES: Use[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”

# Solutions *via Solver*

## *Script*

```
option solver cplex;  
option cplex_options "poolstub=sched poolcapacity=20 \  
    populate=1 poolintensity=4 poolgap=0";  
  
solve;  
  
for {i in 1..Current.npool} {  
    solution ("sched" & i & ".sol");  
    display Work;  
}
```

# Solutions *via Solver*

## *Results*

```
ampl: include schedsPool.run;
CPLEX 12.2.0.2: poolstub=sched
poolcapacity=20
populate=1
poolintensity=4
poolgap=0

CPLEX 12.2.0.2: optimal integer solution; objective 266
464 MIP simplex iterations
26 branch-and-bound nodes

Wrote 20 solutions in solution pool
to files sched1.sol ... sched20.sol.

Solution pool member 1 (of 20); objective 266

Work [*] :=
  1 15    7 14    27 7    70 29    78 29    103 7    115 14
  5 21    11 7    51 7    71 21    87 21    106 38    121 36 ;
```

# Solutions via Solver

## Results (continued)

Solution pool member 2 (of 20); objective 266

Work [\*] :=

```
1 7    5 8    18 7    70 29    78 36    87 14    115 14    121 36
2 28   7 14   65 7    72 7    83 21    106 31    116 7 ;
```

Solution pool member 3 (of 20); objective 266

Work [\*] :=

```
5 21   29 13   51 7    71 34   98 7    115 13
7 15   35 8    64 8    78 16   101 13   116 15
21 7    40 13   70 8    83 8    106 24   121 36 ;
```

Solution pool member 4 (of 20); objective 266

Work [\*] :=

```
2 7    11 7    40 7    71 29    87 15    106 31    121 28
5 22   23 8    64 7    78 13    101 8    115 14    126 7
7 14   29 14   70 14   83 7    102 7    116 7 ;
```

# Solutions *via Solver*: Observations

*Filenames can be formed dynamically*

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

# 3: Heuristic Optimization

*Same model*

*Difficult instances*

- ❖ Set `least_assign` to a “hard” value
- ❖ Get a very good solution quickly

# Heuristic

*Hard case:* least\_assign = 19

```
ampl: model sched1.mod;
ampl: data sched.dat;
ampl: let least_assign := 19;
ampl: option solver cplex;
ampl: solve;

CPLEX 12.2.0.2: optimal integer solution; objective 269
635574195 MIP simplex iterations
86400919 branch-and-bound nodes

ampl: option omit_zero_rows 1, display_1col 0;
ampl: display Work;

Work [*] :=
  4 22    16 39    55 39    78 39    101 39    106 52    122 39
;
```

... *94.8 minutes*



# Heuristic

## *Alternative, indirect approach*

- ❖ Step 1: Relax integrality of **Work** variables  
Solve for zero-one **Use** variables
- ❖ Step 2: Fix **Use** variables  
Solve for integer **Work** variables

*. . . not necessarily optimal, but . . .*

# Heuristic

## *Script*

```
model sched1.mod;
data sched.dat;
let least_assign := 19;

let {j in SCHEDS} Work[j].relax := 1;
solve;

fix {j in SCHEDS} Use[j];
let {j in SCHEDS} Work[j].relax := 0;
solve;
```

# Heuristic

## *Results*

```
ampl: include sched1-fix.run;
```

```
CPLEX 12.2.0.2: optimal integer solution; objective 268.5  
32630436 MIP simplex iterations  
2199508 branch-and-bound nodes
```

```
Work [*] :=
```

1	24	32	19	80	19.5	107	33	126	19.5
3	19	66	19	90	19.5	109	19		
10	19	72	19.5	105	19.5	121	19		

```
CPLEX 12.2.0.2: optimal integer solution; objective 269  
2 MIP simplex iterations  
0 branch-and-bound nodes
```

```
Work [*] :=
```

1	24	10	19	66	19	80	19	105	20	109	19	126	20
3	19	32	19	72	19	90	20	107	33	121	19		

... *2.85 minutes*

# Heuristic: Observations

*Models can be changed dynamically*

- ❖ Retain model-like syntax
- ❖ Execute model-related commands
  - \* `fix {j in SCHEDS} Use[j];`
- ❖ Assign values to properties of model components
  - \* `let {j in SCHEDS} Work[j].relax := 1;`

## 4: Pattern Generation

### *Roll cutting*

- ❖ Min rolls cut (or material wasted)
- ❖ Decide number of each pattern to cut
- ❖ Meet demands for each ordered width

### *Generate cutting patterns*

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

# Pattern Generation

## *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];
```

# Pattern Generation

## *Data*

```
param roll_width := 90 ;  
param: WIDTHS: orders :=  
    60      3  
    30      21  
    25.5    94  
    20      50  
    17.25   288  
    15      178  
    12.75   112  
    10      144 ;
```

# Pattern Generation

## *Script (initialize)*

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

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

let maxPAT := 100000000;

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



# Pattern Generation

## *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;
  }
}
```

# Pattern Generation

*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));
```

# Pattern Generation

## *Results*

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

```
Gurobi 4.6.1: optimal solution; objective 164
```

```
15 simplex iterations
```

```
290 patterns, 164 rolls
```

Cut	3	7	50	44	17	25	2	16
60.00	1	0	0	0	0	0	0	0
30.00	0	3	0	0	0	0	0	0
25.50	0	0	1	1	0	0	0	0
20.00	0	0	0	0	3	0	0	0
17.25	0	0	3	2	0	2	0	0
15.00	2	0	0	2	2	2	0	0
12.75	0	0	1	0	0	2	7	0
10.00	0	0	0	0	0	0	0	9

```
WASTE = 0.32%
```

# Pattern Generation

## *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 ;
```

# Pattern Generation

## *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%
```

# Pattern Generation

## *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 ;
```

# Pattern Generation

## *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%
```

# Pattern Generation: Observations

## *Parameters can serve as script variables*

- ❖ Declare as in model
  - \* `param pattern {WIDTHS} integer >= 0;`
- ❖ Use in algorithm
  - \* `let pattern[curr_width] := pattern[curr_width] - 1;`
- ❖ Assign to model parameters
  - \* `let {w in WIDTHS} nbr[w,nPAT] := pattern[w];`

## *Scripts are easy to modify*

- ❖ Store only every 100<sup>th</sup> pattern found
  - \* `if nPAT mod 100 = 0 then`
    - `let {w in WIDTHS} nbr[w,nPAT/100] := pattern[w];`



# 5: Decomposition

## *Stochastic nonlinear location-transportation*

- ❖ Min expected total cost
  - \* Nonlinear construction costs at origins
  - \* Linear transportation costs from origins to destinations
- ❖ Stochastic demands with recourse
  - \* Decide what to build
  - \* Observe demands and decide what to ship

## *Solve by Benders decomposition*

- ❖ Nonlinear master problem
- ❖ Linear subproblem for each scenario

# Decomposition

*Original model (sets, parameters, variables)*

```
set WHSE;    # shipment origins (warehouses)
set STOR;    # shipment destinations (stores)

param build_cost {WHSE} > 0;    # costs per unit to build warehouse
param build_limit {WHSE} > 0;    # limits on units shipped

var Build {i in WHSE} >= 0, <= .9999 * build_limit[i];
                                     # capacities of warehouses to build

set SCEN;    # demand scenarios

param prob {SCEN} >= 0, <= 1;    # probabilities of scenarios
param demand {STOR,SCEN} >= 0;    # amounts required at stores

param ship_cost {WHSE,STOR} >= 0;    # shipment costs per unit
var Ship {WHSE,STOR,SCEN} >= 0;    # amounts to be shipped
```

# Decomposition

*Original model (objective, constraints)*

```
minimize Total_Cost:
    sum {i in WHSE}
        build_cost[i] * Build[i] / (1 - Build[i]/build_limit[i]) +
    sum {s in SCEN} prob[s] *
        sum {i in WHSE, j in STOR} ship_cost[i,j] * Ship[i,j,s];

subj to Supply {i in WHSE, s in SCEN}:
    sum {j in STOR} Ship[i,j,s] <= Build[i];

subj to Demand {j in STOR, s in SCEN}:
    sum {i in WHSE} Ship[i,j,s] = demand[j,s];
```

# Decomposition

*Sub model (sets, parameters, variables)*

```
set WHSE;    # shipment origins (warehouses)
set STOR;    # shipment destinations (stores)

param build {i in WHSE} >= 0, <= .9999 * build_limit[i];
                                     # capacities of warehouses built

set SCEN;    # demand scenarios

param prob {SCEN} >= 0, <= 1;    # probabilities of scenarios
param demand {STOR,SCEN} >= 0;  # amounts required at stores

param ship_cost {WHSE,STOR} >= 0; # shipment costs per unit
var Ship {WHSE,STOR,SCEN} >= 0;  # amounts to be shipped
```

# Decomposition

*Sub model (objective, constraints)*

```
param S symbolic in SCEN;

minimize Scen_Ship_Cost:
    prob[S] * sum {i in WHSE, j in STOR} ship_cost[i,j] * Ship[i,j];

subj to Supply {i in WHSE}:
    sum {j in STOR} Ship[i,j] <= build[i];

subj to Demand {j in STOR}:
    sum {i in WHSE} Ship[i,j] = demand[j,S];
```

# Decomposition

*Master model (sets, parameters, variables)*

```
param build_cost {WHSE} > 0;      # costs per unit to build warehouse
param build_limit {WHSE} > 0;    # limits on units shipped

var Build {i in WHSE} >= 0, <= .9999 * build_limit[i];
                                # capacities of warehouses to build

param nCUT >= 0 integer;

param cut_type {SCEN,1..nCUT} symbolic
  within {"feas","infeas","none"};

param supply_price {WHSE,SCEN,1..nCUT} <= 0.000001;
param demand_price {STOR,SCEN,1..nCUT};

var Max_Exp_Ship_Cost {SCEN} >= 0;
```

# Decomposition

## *Master model (objective, constraints)*

```
minimize Expected_Total_Cost:
    sum {i in WHSE}
        build_cost[i] * Build[i] / (1 - Build[i]/build_limit[i]) +
    sum {s in SCEN} Max_Exp_Ship_Cost[s];

subj to Cut_Defn {s in SCEN, k in 1..nCUT: cut_type[s,k] != "none"}:
    if cut_type[s,k] = "feas" then Max_Exp_Ship_Cost[s] else 0 >=
        sum {i in WHSE} supply_price[i,s,k] * Build[i] +
        sum {j in STOR} demand_price[j,s,k] * demand[j,s];
```

# Decomposition

## *Script (initialization)*

```
model stbenders.mod;
data stnltrnloc.dat;

suffix dunbdd;
option presolve 0;

problem Sub: Ship, Scen_Ship_Cost, Supply, Demand;
    option solver cplex;
    option cplex_options 'primal presolve 0';

problem Master: Build, Max_Exp_Ship_Cost, Exp_Total_Cost, Cut_Defn;
    option solver minos;

let nCUT := 0;

param GAP default Infinity;
param RELGAP default Infinity;
param Exp_Ship_Cost;
```



# Decomposition

## *Script (iteration)*

```
repeat {  
    solve Master;  
    let {i in WHSE} build[i] := Build[i];  
    let Exp_Ship_Cost := 0;  
    let nCUT := nCUT + 1;  
    for {s in SCEN} {  
        let S := s;  
        solve Sub;  
        ... generate a cut ...  
    }  
    if forall {s in SCEN} cut_type[s,nCUT] != "infeas" then {  
        let GAP := min (GAP,  
            Exp_Ship_Cost - sum {s in SCEN} Max_Exp_Ship_Cost[s]);  
        let RELGAP := 100 * GAP / Expected_Total_Cost;  
    }  
} until RELGAP <= .000001;
```

# Decomposition

## *Script (cut generation)*

```
for {s in SCEN} {
  let S := s;
  solve Sub;

  if Sub.result = "solved" then {
    let Exp_Ship_Cost := Exp_Ship_Cost + Scen_Ship_Cost;

    if Scen_Ship_Cost > Max_Exp_Ship_Cost[s] + 0.00001 then {
      let cut_type[s,nCUT] := "feas";
      let {i in WHSE} supply_price[i,s,nCUT] := Supply[i].dual;
      let {j in STOR} demand_price[j,s,nCUT] := Demand[j].dual;
    }

    else let cut_type[s,nCUT] := "none";
  }

  else if Sub.result = "infeasible" then {
    let cut_type[s,nCUT] := "infeas";
    let {i in WHSE} supply_price[i,s,nCUT] := Supply[i].dunbdd;
    let {j in STOR} demand_price[j,s,nCUT] := Demand[j].dunbdd;
  }
}
```

# Decomposition

## *Results*

```
ampl: include stbenders.run;
MASTER PROBLEM 1: 0.000000
SUB-PROBLEM 1 low: infeasible
SUB-PROBLEM 1 mid: infeasible
SUB-PROBLEM 1 high: infeasible
MASTER PROBLEM 2: 267806.267806
SUB-PROBLEM 2 low: 1235839.514234
SUB-PROBLEM 2 mid: 1030969.048921
SUB-PROBLEM 2 high: infeasible
MASTER PROBLEM 3: 718918.236014
SUB-PROBLEM 3 low: 1019699.661119
SUB-PROBLEM 3 mid: 802846.293052
SUB-PROBLEM 3 high: 695402.974379
GAP = 2517948.928551, RELGAP = 350.241349%
```

# Decomposition

## *Results (continued)*

MASTER PROBLEM 4: 2606868.719958

SUB-PROBLEM 4 low: 1044931.784272

SUB-PROBLEM 4 mid: 885980.640150

SUB-PROBLEM 4 high: 944581.118758

GAP = 749765.716399, RELGAP = 28.761161%

MASTER PROBLEM 5: 2685773.838398

SUB-PROBLEM 5 low: 1028785.052062

SUB-PROBLEM 5 mid: 815428.531237

SUB-PROBLEM 5 high: 753627.189086

GAP = 394642.837091, RELGAP = 14.693822%

MASTER PROBLEM 6: 2743483.001029

SUB-PROBLEM 6 low: 1000336.408156

SUB-PROBLEM 6 mid: 785602.983289

SUB-PROBLEM 6 high: 725635.817601

GAP = 222288.965560, RELGAP = 8.102436%

# Decomposition

## *Results (continued)*

MASTER PROBLEM 7: 2776187.713412

SUB-PROBLEM 7 low: 986337.500000

SUB-PROBLEM 7 mid: 777708.466300

SUB-PROBLEM 7 high: 693342.659287

GAP = 59240.084058, RELGAP = 2.133864%

MASTER PROBLEM 8: 2799319.395374

SUB-PROBLEM 8 low: 991426.284976

SUB-PROBLEM 8 mid: 777146.351060

SUB-PROBLEM 8 high: 704353.854398

GAP = 38198.286498, RELGAP = 1.364556%

MASTER PROBLEM 9: 2814772.778136

SUB-PROBLEM 9 low: 987556.309573

SUB-PROBLEM 9 mid: 772147.258329

SUB-PROBLEM 9 high: 696060.666966

GAP = 17658.226624, RELGAP = 0.627341%

# Decomposition

## *Results (continued)*

```
MASTER PROBLEM 10: 2818991.649514
SUB-PROBLEM 10 mid: 771853.500000
SUB-PROBLEM 10 high: 689709.131427
GAP = 2361.940101, RELGAP = 0.083787%
MASTER PROBLEM 11: 2819338.502316
SUB-PROBLEM 11 high: 692406.351318
GAP = 2361.940101, RELGAP = 0.083776%
MASTER PROBLEM 12: 2819524.204253
SUB-PROBLEM 12 high: 690478.286312
GAP = 541.528304, RELGAP = 0.019206%
MASTER PROBLEM 13: 2819736.994159
GAP = -0.000000, RELGAP = -0.000000%
OPTIMAL SOLUTION FOUND
Expected Cost = 2819736.994159
```

# Decomposition: Observations

## *Loops can iterate over sets*

- ❖ Solve a subproblem for each scenario
  - \* for {s in SCEN} { ...

## *One model can represent all subproblems*

- ❖ Assign loop index s to set S, then solve
  - \* let S := s;
  - solve Sub;

## *Results of solve can be tested*

- ❖ Check whether optimization was successful
  - \* if Sub.result = "solved" then { ...
  - \* else if Sub.result = "infeasible" then { ...

# Concluding 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
- ❖ More database connections
- ❖ IDE for debugging
- ❖ APIs for popular languages (C++, Java, C#, VB, *Python*)