

Update on AMPL Extensions for Stochastic Programming

David M. Gay

AMPL Optimization LLC

dmg@ampl.com

This talk involves work done partly at Sandia National Labs and includes material released as SAND2006-4717C. Sandia is a multiprogram laboratory operated by Sandia Corporation, a Lockheed Martin Company, for the United States Department of Energy's National Nuclear Security Administration under contract DE-AC04-94AL85000.

AMPL summary

AMPL: a language for

mathematical programming problems:

minimize f(x)

s.t.
$$\ell \leq c(x) \leq u$$
,

with $x \in \mathbb{R}^n$ and $c : \mathbb{R}^n \to \mathbb{R}^m$ given algebraically and some x_i discrete.



AMPL goals

- Easy transcription from math (avoid mistakes)
- Explicit indexing (no hidden magic)
- Declare before use (one-pass reading)
- Separate model, data, commands (orthogonality)
- Separate solvers (open solver interface)
- Update entities as needed (lazy evaluation)
- Builtin math. prog. stuff (presolve, reduced costs)
- Aim for large scale nonlinear (sparsity, generality)

Example model: dieti.mod

```
set NUTR; set FOOD;
param cost {FOOD} > 0;
param f_min \{FOOD\} >= 0;
param f_max {j in FOOD} >= f_min[j];
param n_min {NUTR} >= 0;
param n_max {i in NUTR} >= n_min[i];
param amt {NUTR,FOOD} >= 0;
var Buy {j in FOOD} integer >= f_min[j], <= f_max[j];</pre>
minimize Total_Cost:
            sum {j in FOOD} cost[j] * Buy[j];
subject to Diet {i in NUTR}:
   n_min[i] <= sum {j in FOOD} amt[i,j] * Buy[j]
            <= n_max[i];
```

Example data file: diet2a.dat (beginning)

data; set NUTR := A B1 B2 C NA CAL ; set FOOD := BEEF CHK FISH HAM MCH MTL SPG TUR ; cost f_min f_max := param: BEEF 3.19 2 10 CHK 2.59 2 10 FISH 2.29 2 10 2.89 2 HAM 10 1.89 2 10 MCH 2 MTL 1.99 10 SPG 1.99 2 10 2 TUR 2.49 10

Example data file continued: diet2a.dat

```
param:
          n_min
                  n_max :=
           700
                  20000
   Α
                  20000
   C
           700
                  20000
   B1
           700
                  20000
   B2
           700
                  50000
   NA
              0
   CAL
         16000
                  24000 ;
param amt (tr):
             Α
                  C
                       B1
                             B2
                                    NA
                                          CAL :=
                             15
   BEEF
           60
                 20
                       10
                                   938
                                          295
   CHK
            8
                  0
                       20
                             20
                                  2180
                                          770
                       15
                             10
                                   945
                                          440
   FISH
            8
                 10
                                          430
   MAH
           40
                 40
                       35
                             10
                                   278
                                  1182
   MCH
                       15
                             15
                                          315
            15
                 35
                                          400
   MTL
                 30
                       15
                             15
                                   896
           70
   SPG
           25
                       25
                 50
                             15
                                  1329
                                          370
   TUR
                                  1397
           60
                 20
                       15
                             10
                                          450;
   6
```

Example session

```
ampl: model dieti.mod; data diet2a.dat;
ampl: option solver scplex; solve;
CPLEX 11.2.0: optimal integer solution; objective 119.3
12 MIP simplex iterations; 6 branch-and-bound nodes
ampl: display Buy;
Buy [*] :=
BEEF
 CHK 2
FISH 2
 HAM
 MCH
     10
 MTL
      10
 SPG
 TUR
```

Stochastic Programming — Motivation

Data often not known exactly, e.g.,

- prices
- demands
- rainfall
- transit times
- interest rates
- inflation rates

Stochastic Programming Approaches

Approaches include

- Modifying objective: instead of minimizing f(x),
 - \circ minimize E(f(x))
 - \circ minimize $E(f(x)) + \alpha Var(f(x))$
- Modifying constraints: instead of satisfying a constraint exactly,
 - \circ satisfy with probability 1ϵ
 - \circ fail to satisfy with probability ϵ

What's random?

Potentially random entities include

- lower and upper bounds on
 - variables
 - o constraints
- coefficients, e.g.,
 - o costs
 - o returns
 - o rates
- function arguments

AMPL extension: random variables

Debated whether to add "random parameters" or "random variables".

Internally, they act like nonlinear variables, and "random variable" is a conventional term, so **random** in a **var** declaration introduces a random variable:

```
var x random;
```

Declarations may specify a value (with = or default):

```
var y random = Uniform01();
```

or subsequently be assigned:

```
let x := Normal(0,2);
```

Dependent random variables

Dependent random variables may only be declared in var ... = and var ... default declarations:

```
var x random;
var y = x + 1;
```

Random variables may appear as variables in constraint and objective declarations:

```
s.t. Demand: sum {i in A} build[i] >= y;
```

Seeing random variable values

Printing commands see random variables as strings expressing distributions...

```
var x random = Normal01();
        var y = x + Uniform(3,5);
        display x, y;
gives
        x = 'Normal01()'
        y = 'Uniform(3, 5) + x'
```



Sampling random variables

```
display {1..5} (Sample(x), Sample(y));
```

gives

```
: Sample(x) Sample(y) :=
1   1.51898   3.62453
2   -3.65725   2.50557
3   -0.412257   5.4215
4   0.726723   2.89672
5   -0.606458   3.776
;
```

Conventional uses of random functions

Without random, we get ordinary sampling:

```
var x := Uniform(0,10);
        minimize zot: (x - Normal01())^2;
        display x;
        expand zot;
gives
        x = 6.09209
        minimize zot:
                 (x - 1.51898)^2;
```

New builtin functions

New "builtin" functions for solvers to interpret:

- Expected(ξ)
- Moment $(\xi, n), n = 1, 2, 3, ...$
- Percentile (ξ, p) , $0 \le p \le 100$
- Sample(ξ)
- StdDev (ξ)
- ullet Variance (ξ)
- Probability(logical condition)

What happens when?

Stages indicate what happens when.

SMPS convention: Stage = event followed by decision, perhaps with first stage "event" known.

A variable is split into separate copies, one for each realization of its stage (but not of subsequent stages).

For more on SMPS, see

http://myweb.dal.ca/gassmann/smps2.htm

New "system suffix" .stage New reserved suffix .stage, e.g., set A; set Stages; var x {A, s in Stages} suffix stage s; or var x {A, s in Stages}; let {a in A, s in Stages} x[a,s].stage := s;

Example: stochastic diet problem

Buy in two stages; constrain budget in first stage, suffer random price changes in second stage.

What to buy in first stage?

```
New: set T = 1 .. 2;  # times (stages)
    var Buy {FOOD, t in T} integer >= 0
        suffix stage t;
    s.t. FoodBounds {j in FOOD}: f_min[j]
        <= sum{t in T} Buy[j,t] <= f_max[j];</pre>
```

Stochastic diet problem (cont'd)

```
minimize Total Cost:
             sum {j in FOOD} cost[j] * Buy[j];
New:
        var CostAdj {FOOD} random;
        minimize Total_Cost:
            sum {j in FOOD} cost[j] * Buy[j,1]
            + Expected(sum {j in FOOD}
              cost[j]*CostAdj[j]*Buy[j,2]);
```

Stochastic diet problem (cont'd)

```
sum {j in FOOD} amt[i,j] * Buy[j]
New:
        sum {j in FOOD, t in T}
                  amt[i,j] * Buy[j,t]
        param init_budget;
        s.t. Init_Bud: sum {j in FOOD} Buy[j,1]
                  <= init_budget;
        . . .
        let{j in FOOD} CostAdj[j]
                   := Uniform(.7, 1.3);
```

"Constant" distributions

Assign numerical value to random variable \Longrightarrow simplified problem (for debugging and model development).

Example:

```
let{j in FOOD} CostAdj[j]
:= Sample(Uniform(.7, 1.3));
```

With imported function Expected(x) = x, this works with conventional solvers.

Some things work now

Things that work include

- Most details of random-variable handling
 - Declarations
 - Assignments of distributions
 - Assignments of constants
 - Printing and sampling (in AMPL sessions)
 - Determining what the solver will see as linear
- Writing .nl files with random distributions
- Suffix ".stage" and functions of distributions.

Nonanticipitivity

Nonanticipitivity is implicit in stating problems (compact form). The .nl file has sparsity structure for all constraints and objectives, indicating which variables appear (and giving linear coefficients). This includes random variables. Stage structure is in .stage suffixes. Solvers can split variables if desired.

Work in progress

Updates to solver-interface library (for sampling), sample drivers not yet finished. Plans include

- Routines to pose deterministic equivalents, e.g., with stratified sampling such as Latin hypercube. Options randoptions and (\$solver)_randoptions would control sampling and discretization.
- Program to write .nl file for deterministic equivalent.
- Program to write SMPS format.
- Solver drivers, e.g., for Gassmann's MSLiP.

Bound computations

Forthcoming additions to ASL (AMPL/Solver interface Library) include routines for bound computations. See paper "Bounds from Slopes" in

http://www.sandia.gov/~dmgay/bounds10.pdf

Possible application: importance sampling. Sample next where support measure times variation bound is largest.

For more details (dmg@ampl.com)

http://www.ampl.com points to

- The AMPL book
- examples (models, data)
- descriptions of new stuff (in book 2nd ed., not 1st)
- downloads
 - o student binaries; trial-license form
 - o solver interface library source
 - "standard" table handler & source
 - o papers and reports