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Overview of and Update on ASL, the AMPL/Solver Interface Library

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AMPL: a language for
mathematical programming problems:

$$\begin{aligned} & \text{minimize } f(x) \\ & \text{s.t. } \ell \leq c(x) \leq u, \end{aligned}$$

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



Sample session

```
ampl: model dieti.mod;
ampl: data diet2a.dat;
ampl: option solver gurobi;
ampl: solve;
Gurobi 4.0.0: optimal solution; objective 119.3
28 simplex iterations
21 branch-and-cut nodes
absmipgap = 2.84e-14, relmipgap = 2.38e-16
ampl: display Buy;
Buy [*] :=
BEEF    9
    CHK    2
```



FISH 2

HAM 8

MCH 10

MTL 10

SPG 7

TUR 2

;



What did “solve;” do?

- Write temp. “.nl” file representing the problem;
- invoke `$solver`, passing current *environment* — option names and values;
- read solution from “.sol” file.

Solver uses AMPL/Solver interface Library (ASL) to

- obtain problem details;
- do nonlinear evaluations (if needed);
- return solution — write “.sol” file.



Problem details in .nl file

A .nl file conveys

- Problem dimensions: e.g., numbers of
 - continuous and discrete variables
 - linear and nonlinear constraints and objectives
- initial primal and dual variable values
- initial basis
- *expression graphs* for nonlinearities.



Readers of .nl files

ASL offers several `.nl` readers that prepare for solving various kinds of problems, including

- linear
- quadratic
- nonlinear functions and gradients
- nonlinear functions, gradients, and Hessians:
 - Hessian-vector products
 - explicit (sparse) Hessians
 - exploiting (or ignoring) partial separability.

Suffixes



Suffixes convey auxiliary information on variables, constraints, objectives, problems.

Most commonly, `.sstatus` (for “solver status”) supplies initial basis values:

```
AMPL: option sstatus_table;  
option sstatus_table '\  
0      none      no status assigned\  
1      bas       basic\  
2      sup       superbasic\  
3      low       nonbasic <= (normally =) lower bound\  
4      upp       nonbasic >= (normally =) upper bound\  
5      equ       nonbasic at equal lower and upper bounds\  
6      btw       nonbasic between bounds\  
';
```


.astatus



Suffix `.astatus` (for “AMPL status”) gives AMPL’s view of constraints and variables:

```
AMPL: option astatus_table;
option astatus_table '\
0      in      normal state (in problem)\
1      drop    removed by drop command\
2      pre     eliminated by presolve\
3      fix     fixed by fix command\
4      sub     defined variable, substituted out\
5      unused  not used in current problem\
6      log     logical constraint in current problem\
7      rand    random variable in current problem';
```

.status



Suffix `.status` is `.sstatus` if `.astatus` is “in” and is `.astatus` otherwise:

```
ampl: option solver minos; solve;
MINOS 5.51: ignoring integrality of 8 variables
MINOS 5.51: optimal solution found.
3 iterations, objective 118.0594032
ampl: display Buy, Buy.astatus, Buy.sstatus, Buy.status;
:      Buy      Buy.astatus Buy.sstatus Buy.status      :=
BEEF   5.36061   in          bas          bas
CHK    2         in          low         low
FISH   2         in          low         low
HAM    10        in          upp         upp
MCH    10        in          upp         upp
MTL    10        in          upp         upp
SPG    9.30605   in          bas          bas
```



Suffixes for SOS sets

Special Ordered Set (SOS) structure can be supplied explicitly with `.sosno` and `.ref`.

Most commonly, SOS2 sets are used to express piecewise-linear terms.

AMPL renders nonconvex

<<bkpoint-list ; slope-list>> var

terms with extra constraints and variables adorned with `.sos` and `.sosref` suffixes.

Interfaces to solvers that handle SOS sets can call the ASL's `suf_sos(...)` to remove the added constraints and variables that imply SOS conditions.



Returned suffixes

Solver interfaces can return new or updated suffix values via the `.sol` file.

Many solvers return updated `.sstatus` values.

For unbounded problems, some return `.unbdd` (for a feasible ray). For unbounded dual problems, some return `.dunbdd`.

For MIPs, some return `.absmipgap` or `.relmipgap` and `.bestbound` or `.bestnode` on objectives or problems.



Example of .dunbdd

```
ampl: reset; model diet.mod; data diet2.dat;
ampl: option solver cplex; solve;
CPLEX 12.2.0.0: No LP presolve or aggregator reductions.
infeasible problem.
2 dual simplex iterations (0 in phase I)
constraint.dunbdd returned
suffix dunbdd OUT;
ampl: display _conname, _con.dunbdd;
:      _conname      _con.dunbdd      :=
1      "Diet['A']"      0
2      "Diet['B1']"      0
3      "Diet['B2']"      0.226598
4      "Diet['C']"      0
5      "Diet['NA']"      -0.00255754
6      "Diet['CAL']"      0
```



Forthcoming ASL enhancements

- Support for stochastic programming, including new `.stage` suffix.
- More constraint programming facilities: e.g., variables in subscripts.
- New routines for bounds from slopes.
- New suffix for multi-level program structure.
- Updated documentation.



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:

- $\text{Expected}(\xi)$
- $\text{Moment}(\xi, n), n = 1, 2, 3, \dots$
- $\text{Percentile}(\xi, p), 0 \leq p \leq 100$
- $\text{Sample}(\xi)$
- $\text{StdDev}(\xi)$
- $\text{Variance}(\xi)$
- $\text{Probability}(\textit{logical condition})$



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?

Old: `var Buy {j in FOOD} integer >= f_min[j],
 <= f_max[j];`

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];`



Stochastic diet problem (cont'd)

Old:

```
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)

Old: $\text{sum } \{j \text{ in FOOD}\} \text{ amt}[i,j] * \text{Buy}[j]$

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



“Constant” distributions

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

Example:

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

With imported function $\text{Expected}(x) = x$, 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.



Nonanticipativity

Nonanticipativity 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
 - student binaries; trial-license form
 - solver interface library source
 - “standard” table handler & source
 - papers and reports



Aux. slide: 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];
```

```
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];
```




Aux. slide: diet2a.dat

```
param:    cost    f_min    f_max    :=  
    BEEF    3.19    2        10  
    CHK     2.59    2        10  
    FISH    2.29    2        10  
    HAM     2.89    2        10  
    MCH     1.89    2        10  
    MTL     1.99    2        10  
    SPG     1.99    2        10  
    TUR     2.49    2        10    ;
```

```
param:    n_min    n_max    :=  
    A        700    20000  
    C        700    20000  
    B1       700    20000
```



Aux. slide: diet2a.dat *cont'd*

```
B2      700    20000
NA       0    50000
CAL  16000    24000 ;
```

param amt (tr):

	A	C	B1	B2	NA	CAL	:=
BEEF	60	20	10	15	938	295	
CHK	8	0	20	20	2180	770	
FISH	8	10	15	10	945	440	
HAM	40	40	35	10	278	430	
MCH	15	35	15	15	1182	315	
MTL	70	30	15	15	896	400	
SPG	25	50	25	15	1329	370	
TUR	60	20	15	10	1397	450	;