

\$TITLE: M5-2.GMS, example of non-linear, constrained least squares

\$ONTEXT

there are I observations on two variables (set J), a dependent variables Y and an independent variable J.

the objective is to estimate a log linear relationship via OLS minimizing the sum of squared deviations

imagine that this is production data (two inputs), added constraint equation imposes constant returns to scale: sum of slope coeff = 1

Y = output, X1 = capital, X2 = labor

\$OFFTEXT

```
SETS  I      observations           /I1*I6/  
        J      dep and ind var       /J1*J3/  
        K(J)   set of independent variables /J2*J3/;
```

PARAMETERS

```
Y0(I)  
X0(I,K);
```

TABLE BENCH(I,J)

	J1	J2	J3
I1	4	2	2
I2	3	4	1
I3	10	6	5
I4	14	11	3
I5	18	13	4
I6	22	14	6;

DISPLAY BENCH;

Y0(I) = BENCH(I, "J1");
X0(I,K) = BENCH(I, K);

DISPLAY Y0, X0;

VARIABLES

ALPHA intercept
BETA(K) slope coefficients (elasticities since estimated in logs)
DEV sum of squared deviations
YHAT(I) fitted values of the dependent variable;

EQUATIONS

OBJECTIVE objective function = sum of squared residuals
EYHAT(I) equation for the fitted values of Y (log linear)
CRS constraint constant returns: sum of slope coefficients = 1;

```
OBJECTIVE..  DEV =E= SUM(I, (YHAT(I) - Y0(I))*(YHAT(I) - Y0(I)));
EYHAT(I)..  LOG(YHAT(I)) =E= ALPHA + SUM(K, BETA(K)*LOG(X0(I,K)));
CRS..       SUM(K, BETA(K)) =E= 1;
```

** model OLS: unconstrained OLS*

```
MODEL OLS /OBJECTIVE, EYHAT/;
```

```
ALPHA.L    = 1;
BETA.L(K)  = 1;
YHAT.L(I)  = 2;
```

```
SOLVE OLS USING NLP MINIMIZING DEV;
```

** model OLSC: constrained least squares, imposes CRS*

```
MODEL OLSC /ALL/;
```

```
SOLVE OLSC USING NLP MINIMIZING DEV;
```

** process output to get observed and fitted values of Y*

PARAMETER

RESULTS(I, *);

RESULTS(I, "YHAT") = YHAT.L(I);

RESULTS(I, "Y0") = Y0(I);

DISPLAY RESULTS;

\$LIBINCLUDE XLDUMP RESULTS M5.XLS SHEET1!B3