

```

new ;
closeall ;

// *****
//
// incae_tutorial_03_solution_dynamic_game.gss
//
// Program for the solution of dynamic game of market entry-exit with 2 firms
//
// Given parameters fixed by user, the program:
//     (a) computes an equilibrium (Markov Perfect equilibrium) of the model
//
// by Victor Aguirregabiria
//
// *****

library pgraph ;

// -----
// OUTPUT FILE
// -----
wdir = "C:\\Users\\victo\\Dropbox\\documents\\COURSES\\2022_INCAE\\TUTORIALS\\" ;
// Working directory
buff = changedir(wdir) ;
fileout = "incae_tutorial_03_solution_dynamic_game.out" ; // Name of output file
output file = ^fileout reset ;
format /mb1 /ros 16,4 ;

// -----
// Specification of Profit Function
//
// i = index for firm. i belongs to {1, 2}
// Firms' actions = a_1 and a_2
//
//     a_i = 0 ----> Firm i is not active in market
//     a_i = 1 ----> Firm i is active in market
//
// -----
//
// The game is dynamic because there is a sunk cost of entry
//
// The only state variables of this game are the indicators of whether
// the firms were active or not in the market at previous period.
// These are payoff relevant state variables because they determine
// whether a firm has to pay an entry cost to operate in the market.
//
//     x_i = a_i[t-1]
//
//     x_i = 0 ----> Firm i was not active in market at previous period
//     = 1 ----> Firm i was active in market at previous period

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//
// -----
//
// Parameters:
//     ecost =   Entry cost
//     pi_mon =   Operating profit of a monopolist
//     pi_duo =   Operating profit of a duopolist
//
// -----
//
// Profit function for firm 1 (the symmetric for firm 2)
//
//     Profit_1(a_1 = 0) = eps_1(0)
//
//     Profit_1(a_1 = 1) = (1-a_2) * pi_mon + a_2 * pi_duo
//
//                          - (1-a_1[t-1]) * ecost + eps_1(1)
//
// where eps_1(0) and eps_1(1) are private information shocks that are
// i.i.d. over time and over players, and independent of each other,
// with Standard Normal distribution.
//
// -----

// -----
// 1. Values of parameters and other constants
// -----
ecost = 1.0 ;    // Entry cost
pi_mon = 1.0 ;   // Monopoly profit
pi_duo = 0.4 ;   // Duopoly profit

dfact = 0.9 ;    // Discount factor

nplayer = 2 ;    // Number of players

trueparam = (ecost | pi_mon | pi_duo) ; // vector with true values of parameters
namesb = "Entry Cost"
        | "Monopoly Prof"
        | "Duopoly Prof" ; // Vector with names of parameters
kparam = rows(trueparam) ; // Number of parameters to estimate

// 'vstate' is a matrix with all the possible values of
// the state variables of this dynamic game.
// That is, a matrix with all the possible values of x_1 and x_2
//
// - Each row represents a value for the state variables
// - Column 1 contains the values of x_1, and column 2 the values of x_2

vstate = (0 ~ 0)
        | (0 ~ 1)

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| (1 ~ 0)
| (1 ~ 1) ;

nstate = rows(vstate) ; // Number of possible values of the vector of state
variables

//
-----
-----
// 2. PROCEDURE FOR EVALUATING BEST-RESPONSE MAPPING
//
-----
-----
//
// Input of the procedure:
//     prob0 = (nstate x 2) matrix with the CCPs of each player
//
//     Each row corresponds to a state, each column corresponds to the CCP of a
//     player
//
// Output of the procedure:
//     newprob = (nstate x 2) matrix with the best response CCPs of each player
//
//     The i-th column is the vector of CCPs which are player i's best response
//     to prob0
//
//
-----
-----

proc (1) = equilmap(prob0);
    local ns, newprob, profit1_a0, profit1_a1, profit2_a0, profit2_a1,
        iptran1_a0, iptran1_a1, iptran2_a0, iptran2_a1,
        eprofit1, eprofit2, ftran, value1, value2, vtilde1, vtilde2 ;
    ns = rows(prob0) ; // number of states
    newprob = prob0 ;

    // -----
    // a. Vectors of expected profits
    // -----

    // Expected profit of firms 1 and 2 if own action is 0
    profit1_a0 = zeros(ns,1) ;
    profit2_a0 = zeros(ns,1) ;

    // Expected profit of firms 1 and 2 if own action is 1
    profit1_a1 = (1-prob0[:,2]).*pi_mon + prob0[:,2].*pi_duo -
ecost.*(1-vstate[:,1]) ;
    profit2_a1 = (1-prob0[:,1]).*pi_mon + prob0[:,1].*pi_duo -

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ecost.*(1-vstate[:,2]) ;

// Ex-ante expected profit of firms 1 and before knowing the own epsilons
// Assumption about distribution of epsilon: Standard Normal
eprofit1 = (1-prob0[:,1]).*profit1_a0 + prob0[:,1].*profit1_a1
          + sqrt(2) * pdfn(cdfni(prob0[:,1])) ;

eprofit2 = (1-prob0[:,2]).*profit2_a0 + prob0[:,2].*profit2_a1
          + sqrt(2) * pdfn(cdfni(prob0[:,2])) ;

// -----
// b. Transition probabilities of the state variables
// -----
//
// - (nstate x nstate) matrices of transition probabilities of
//   state variables
//
// - In these transition matrices:
//     - Rows represent current value of state variables
//     - Columns represent next period value of state variables
//       (the elements of each row should sum to 1)
//
// - Remember: vstate = (0~0) | (0~1) | (1~0) | (1~1)
//
// - Since the only state variables are the lagged values
//   of firms' actions, all what we need to obtain the
//   transition probabilities is the choice probabilities
//
// - Transition matrices from the point of view of each firm:
//   Each firm knows his current action but not the current action
//   of his competitor
//
// -----

// Transition matrix of firm 1 if current own action = 0
// Remember: vstate = (0~0) | (0~1) | (1~0) | (1~1)
iptran1_a0 = (1-prob0[:,2]) ~ prob0[:,2] ~ zeros(ns,1) ~ zeros(ns,1) ;

// Transition matrix of firm 1 if current own action = 1
// Remember: vstate = (0~0) | (0~1) | (1~0) | (1~1)
iptran1_a1 = zeros(ns,1) ~ zeros(ns,1) ~ (1-prob0[:,2]) ~ prob0[:,2] ;

// Transition matrix of firm 2 if current own action = 0
// Remember: vstate = (0~0) | (0~1) | (1~0) | (1~1)
iptran2_a0 = (1-prob0[:,1]) ~ zeros(ns,1) ~ prob0[:,1] ~ zeros(ns,1) ;

// Transition matrix of firm 2 if current own action = 1
// Remember: vstate = (0~0) | (0~1) | (1~0) | (1~1)
iptran2_a1 = zeros(ns,1) ~ (1-prob0[:,1]) ~ zeros(ns,1) ~ prob0[:,1] ;

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// Ex-ante transition probability matrix
//   Not conditional on own action
//   Remember: vstate = (0~0) | (0~1) | (1~0) | (1~1)

ftran = (1-prob0[:,1]).*(1-prob0[:,2])
      ~ (1-prob0[:,1]).*prob0[:,2]
      ~ prob0[:,1].*(1-prob0[:,2])
      ~ prob0[:,1].*prob0[:,2] ;

// -----
// c. Calculating present values for each firm and state
//   Given that firms' believe that they will behave
//   in the future according to the probabilities prob0
// -----
ftran = inv(eye(ns)-dfact*ftran) ; // 'Discounting' matrix
value1 = ftran * eprofit1 ; // Present values for firm 1
value2 = ftran * eprofit2 ; // Present values for firm 2

// Threshold value for firm 1:
//   Value of action 1 minus value of action 0
vtilda1 = (profit1_a1 + dfact * iptran1_a1 * value1)
          - (profit1_a0 + dfact * iptran1_a0 * value1) ;

// Threshold value for firm 2:
//   Value of action 1 minus value of action 0
vtilda2 = (profit2_a1 + dfact * iptran2_a1 * value2)
          - (profit2_a0 + dfact * iptran2_a0 * value2) ;

// -----
// d. Best response probabilities
// -----
newprob = cdfn(vtilda1/sqrt(2)) ~ cdfn(vtilda2/sqrt(2)) ;
retp(newprob) ;
endp ;
// ----- End procedure equimap -----

// -----
// 3. COMPUTING AN EQUILIBRIUM (MPE)
// -----
//
// Here the algorithm for computing an equilibrium
// is fixed point iterations in the best response equilibrium mapping
//
// -----

prob0 = rndu(nstate,nplayer) ; // Initial value for the vector of players' CCPs

cconv = 1e-12 ; // Convergence constant
criter = 1000 ; // Initial value for the convergence criterion

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maxiter = 100 ; // Maximum number of iterations to stop
                // Remember, the equilibrium mapping is not a contraction
                // such that fixed point iterations might not converge

iter=1 ;
do while (criter>cconv).and(iter<=maxiter) ;
    "          Best response mapping iteration  =";; iter ;
    "          Convergence criterion =";; criter ;
    "" ;
    prob1 = equilmap(prob0);
    criter = maxc(maxc(abs(prob1-prob0))) ;
    prob0 = prob1 ;
    iter=iter+1 ;
enddo ;
pequil = prob0 ;

if (iter<maxiter) ;
    "-----" ;
    "          CONVERGENCE ACHIEVED AFTER";; iter;; "BEST RESPONSE ITERATIONS";
    "-----" ;
    "          EQUILIBRIUM PROBABILITIES" ;
    pequil ;
    "" ;
    "          Remember that" ;
    "" ;
    "          vstate  = (0 ~ 0)" ;
    "                   | (0 ~ 1)" ;
    "                   | (1 ~ 0)" ;
    "                   | (1 ~ 1)" ;
    "-----" ;
endif ;

// -----
// 4. COMPUTING THE STEADY-STATE DISTRIBUTION OF THE STATE VARIABLES
//     for THE COMPUTED EQUILIBRIUM
//
//     This is the distribution of "market structure" in steady state
// -----

// Transition probability matrix for the state variables
ftran = (1-pequil[.,1]).*(1-pequil[.,2])
        ~ (1-pequil[.,1]).*pequil[.,2]
        ~ pequil[.,1].*(1-pequil[.,2])
        ~ pequil[.,1].*pequil[.,2] ;

// By definition, the steady-state distribution 'psteady'
// is such that:
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//      psteady = Ftran * psteady
//
//  where Ftran is the transition probability matrix of the state vars
//
//  We can calculate psteady by fixed point iterations in the vector
//  system:  psteady = Ftran * psteady
//  -----

cconv = 1e-6 ;
criter = 1000 ;
psteady = (1/nstate)*ones(nstate,1) ;
do while criter>cconv ;
    "Criter =";; criter ;
    pbuff = ftran'*psteady ;
    criter = maxc(abs(pbuff-psteady)) ;
    psteady = pbuff ;
endo ;

"-----" ;
"      Ergodic Distribution of the State Variables" ;
"" ;
psteady ;

"" ;
"-----" ;
"" ;
"      Remember that" ;
"" ;
"          vstate = (0 ~ 0)" ;
"                  | (0 ~ 1)" ;
"                  | (1 ~ 0)" ;
"                  | (1 ~ 1)" ;
"" ;
"-----" ;

output off ;

end ;

```