

```

new ;
closeall ;

// ****
// 
// incae_tutorial_04_estimation_dynamic_game.gss
//
// Program for the estimation 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
//   (b) generates an artificial / simulated dataset from that equilibrium
//   (c) uses the simulated as actual data and estimates the parameters of the
model
//
// The estimation methods are:
//   (a) Two-step Pseudo Maximum Likelihood
//   (b) Iterated Pseudo Maximum Likelihood (NPL) (Full solution method)
//
// by Victor Aguirregabiria
//
// ****

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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.

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// 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
//
// -----
//
// 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
// -----
nobs = 1000 ; // Number of observations (local markets) in simulated dataset
npliter = 20 ; // Maximum Number of NPL iterations for estimation

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

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// '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)
| (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, vtilda1, vtilda2 ;
  ns = rows(prob0) ;    // number of states
  newprob = prob0 ;

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

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// -----
// 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 -
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
+ pdfn(cdfni(prob0[.,1])) ;

eprofit2 = (1-prob0[.,2]).*profit2_a0 + prob0[.,2].*profit2_a1
+ 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
//
// -----
// Trasition 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) ;

// Trasition matrix of firm 1 if current own action = 1
// Remember: vstate = (0~0) | (0~1) | (1~0) | (1~1)

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iptran1_a1 = zeros(ns,1) ~ zeros(ns,1) ~ (1-prob0[.,2]) ~ prob0[.,2] ;

// Trasition 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) ;

// Trasition 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] ;

// 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) ~ cdfn(vtilda2) ;
retp(newprob) ;
endp ;
// ----- End procedure equimap -----

// -----
// 3. COMPUTING AN EQUILIBRIUM (MPE)
// -----

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// -----
// 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
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 ;
endo ;
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
// -----

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// 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:
//
//      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)" ;
" ;
"-----" ;

// -----
// 5. PROCEDURE TO SIMULATE DATA FROM THE COMPUTED EQUILIBRIUM
// -----
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// Inputs of the procedure:
//
//      numobs = Number of observations (markets) in the simulated dataset
//      pchoice = Vector of equilibrium CCPs
//      pste    = Vector with steady-state distribution of the state variables
//      vecs    = Matrix with the values of the possible values of the state
variables
//          (in a two-player model of market entry-exit, we have that
//           vecs = (0 ~ 0) | (0 ~ 1) | (1 ~ 0) | (1 ~ 1)
//
// Outputs of the procedure:
//      simul_a     = (numobs x 2) matrix with observations at every market
//                      of actions of firms 1 and 2 at period t
//      simul_a_1   = (numobs x 2) matrix with observations at every market
//                      of actions of firms 1 and 2 at period t-1
//
// -----
proc (2) = simdygam(numobs,pchoice,pste,vecs) ;
  local seed, nums, nump, pbuff0, pbuff1, uobs, simul_a, simul_a_1 ;
  nums = rows(vecs) ;
  nump = cols(pchoice) ;
  pbuff1 = cumsumc(pste) ;
  pbuff0 = cumsumc((0|pste[1:nums-1])) ;
  uobs = rndu(numobs,1) ;
  uobs = ((uobs.>=(pbuff0')).*(uobs.<=(pbuff1')))* seqa(1,1,nums) ;
  simul_a_1 = vecs[uobs,.] ;
  simul_a = (rndu(numobs,nump).<=pchoice[uobs,.]) ;
  retp(simul_a,simul_a_1) ;
endp ;

// -----
// 6. SIMULATING DATA
// -----
{aobs , aobs_1} = simdygam(nobs,pequil,psteady,vstate) ;

// -----
// 7. PROCEDURE for FREQUENCY ESTIMATOR
// -----
proc (1) = freqprob(yobs,xobs,xval) ;
  local numx, numq, prob1, t, selx, denom, numer ;
  numx = rows(xval) ;
  numq = cols(yobs) ;
  prob1 = zeros(numx,numq) ;
  t=1 ;
  do while t<=numx ;
    selx = prodc((xobs==xval[t,.])') ;
    denom = sumc(selx) ;

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if (denom==0) ;
prob1[t,.] = zeros(1,numq) ;
else ;
numer = sumc(selx.*yobs) ;
prob1[t,.] = (numer')./denom ;
endif ;
t=t+1 ;
endo ;
retp(prob1) ;
endp ;

// -----
// 8. FREQUENCY ESTIMATES OF CCPs
// -----
freq_est_prob = freqprob(aobs,aobs_1,vstate) ;

"" ;
"-----" ;
"      Estimated Ergodic Distribution of State Variables" ;
"" ;
est_psteady = zeros(nstate,1) ;
est_psteady[1] = sumc(prodc((aobs_1==vstate[1,.])'))/nobs ;
est_psteady[2] = sumc(prodc((aobs_1==vstate[2,.])'))/nobs ;
est_psteady[3] = sumc(prodc((aobs_1==vstate[3,.])'))/nobs ;
est_psteady[4] = sumc(prodc((aobs_1==vstate[4,.])'))/nobs ;

est_psteady ;
"" ;
"      True Ergodic Distribution of State Variables" ;
psteady ;
"" ;
"-----" ;
"" ;
"      Estimated Conditional Choice Probabilities" ;
freq_est_prob ;
"" ;
"      True Conditional Choice Probabilities" ;
pequil ;
"" ;
"-----" ;

// -----
// 9. PROCEDURE for CONSTRAINED PROBIT ESTIMATOR
// -----


proc (3) = miprobit(ydum,x,rest,b0,out) ;
local myzero, nobs, nparam, eps, namesb, iter, llike,
      criter, Fxb0, phixb0, lamdab0, dlogLb0,
      d2logLb0, b1, lamda0, lamda1, Avarb, sdb, tstat,

```

```

numy1, numy0, logL0, LRI, pseudoR2, k ;
myzero = 1e-36 ;
nobs = rows(ydum) ;
nparam = cols(x) ;
eps = 1E-6 ;
namesb = seqa(1,1,nparam) ;
namesb = 0 $+ "b" $+ ftocv(namesb,2,0);
iter=1 ;
llike = 1000 ;
criter = 1000 ;
do while (criter>eps) ;
if (out==1) ;
  "" ;
  "Iteration           = " iter ;
  "Log-Likelihood function = " llike ;
  "Criterion          = " criter ;
  "" ;
endif ;
Fxb0 = cdfn(x*b0+rest) ;
Fxb0 = Fxb0 + (myzero - Fxb0).*(Fxb0.<myzero)
      + (1-myzero - Fxb0).*(Fxb0.>(1-myzero));
llike = ydum'*ln(Fxb0) + (1-ydum)'*ln(1-Fxb0) ;
phixb0 = pdfn(x*b0+rest) ;
lamdab0 = ydum.*(phixb0./Fxb0) + (1-ydum).*(-phixb0./(1-Fxb0)) ;
dlogLb0 = x'*lamdab0 ;
d2logLb0 = -((lamdab0.*(lamdab0 + x*b0 + rest)).*x)'*x ;
b1 = b0 - inv(d2logLb0)*dlogLb0 ;
criter = maxc(abs(b1-b0)) ;
b0 = b1 ;
iter = iter + 1 ;
endo ;
Fxb0 = cdfn(x*b0 + rest) ;
Fxb0 = Fxb0 + (myzero - Fxb0).*(Fxb0.<myzero)
      + (1-myzero - Fxb0).*(Fxb0.>(1-myzero));
llike = ydum'*ln(Fxb0) + (1-ydum)'*ln(1-Fxb0) ;
phixb0 = pdfn(x*b0 + rest) ;
lamda0 = -phixb0./(1-Fxb0) ;
lamda1 = phixb0./Fxb0 ;
Avarb = ((lamda0.*lamda1).*x)'*x ;
Avarb = inv(-Avarb) ;
sdb = sqrt(diag(Avarb)) ;
tstat = b0./sdb ;
numy1 = sumc(ydum) ;
numy0 = nobs - numy1 ;
logL0 = numy1*ln(numy1) + numy0*ln(numy0) - nobs*ln(nobs) ;
LRI = 1 - llike/logL0 ;
pseudoR2 = 1 - ( (ydum - Fxb0)'*(ydum - Fxb0) )/numy1 ;
if (out==1) ;
  "Number of Iterations      = " iter ;
  "Log-Likelihood function  = " llike ;

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"Likelihood Ratio Index    = " LRI ;
"Pseudo-R2                 = " pseudoR2 ;
"" ;
"-----";
"      Parameter      Estimate      Standard      t-ratios";
"                           Errors" ;
"-----";
k=1;
do while k<=nparam;
  print $namesb[k];;b0[k];;sdb[k];;tstat[k];
  k=k+1 ;
endo;
"-----";
endif ;
retp(b0,Avarb,llike) ;
endp ;

// -----
// 10. PROCEDURE for NPL ITERATIONS
// -----
proc (3) = npldygam(yobs,yobs_1,pchoice,disfact,mstate,b0,kiter,namesb);
  local myzero, nobs, nplayer, numx, kparam, best, varb,
        iter, cconv, criter, pchoice_0, pchoice_1,
        indobs, u0, u1, e0, e1, ptran, inv_bf,
        iptran1_a0, iptran1_a1, iptran2_a0, iptran2_a1,
        umat1_a0, umat1_a1, umat2_a0, umat2_a1,
        zmat1, zmat2, emat1, emat2,
        zobs, eobs, tetaest, varest, likelihood,
        sdb, tstat, k ;

  // -----
  // Some constants
  // -----
  myzero = 1e-16 ;
  nobs = rows(yobs) ;
  nplayer = cols(yobs) ;
  numx = rows(pchoice) ;
  kparam = rows(b0) ;
  best = zeros(kparam,kiter) ;
  varb = zeros(kparam,kparam*kiter) ;
  yobs = yobs[.,1] | yobs[.,2] ;

  // -----
  // Vector with indexes for the observed state
  // -----
  indobs = 1.*prodc((yobs_1==mstate[1,.])')
    + 2.*prodc((yobs_1==mstate[2,.])')
    + 3.*prodc((yobs_1==mstate[3,.])')
    + 4.*prodc((yobs_1==mstate[4,.])') ;

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```

// -----
// NPL algorithm
// -----
cconv = 1e-6 ;      // convergence constant
criter = 1000 ;    // convergence criterion
pchoice_0 = pchoice ; // Initial vector of CCPs
iter=1 ;
do while (iter<=kiter).and(criter>cconv) ;
// -----
// a. Truncation of probabilities to avoid
//     inverse Mill's ratio = +INF
// -----
pchoice_0[.,1] = (pchoice_0[.,1].<myzero).*myzero
                + (pchoice_0[.,1].>(1-myzero)).*(1-myzero)
                +
(pchoice_0[.,1].>=myzero).*(pchoice_0[.,1].<=(1-myzero)).*pchoice_0[.,1] ;
pchoice_0[.,2] = (pchoice_0[.,2].<myzero).*myzero
                + (pchoice_0[.,2].>(1-myzero)).*(1-myzero)
                +
(pchoice_0[.,2].>=myzero).*(pchoice_0[.,2].<=(1-myzero)).*pchoice_0[.,2] ;

// -----
// b. Matrix of transition probabilities
// -----
ptran = (1-pchoice_0[.,1]).*(1-pchoice_0[.,2])
        ~ (1-pchoice_0[.,1]).*pchoice_0[.,2]
        ~ pchoice_0[.,1].*(1-pchoice_0[.,2])
        ~ pchoice_0[.,1].*pchoice_0[.,2] ;

// -----
// c. Inverse of I-b*F
// -----
inv_bf = inv(eye(numx)-disfact*ptran) ;

// -----
// d. Matrices Pr(a[t] | a[t-1], ai[t])
// -----
iptran1_a0 = (1-pchoice_0[.,2]) ~ pchoice_0[.,2] ~ zeros(nstate,1) ~
zeros(nstate,1) ;
iptran1_a1 = zeros(nstate,1) ~ zeros(nstate,1) ~ (1-pchoice_0[.,2])~
pchoice_0[.,2] ;
iptran2_a0 = (1-pchoice_0[.,1]) ~ zeros(nstate,1) ~ pchoice_0[.,1] ~
zeros(nstate,1) ;
iptran2_a1 = zeros(nstate,1) ~ (1-pchoice_0[.,1])~ zeros(nstate,1) ~
pchoice_0[.,1] ;

// -----
// e. Construction of explanatory variables
// -----

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umat1_a0 = zeros(numx,kparam) ;
umat1_a1 = (-(1-mstate[.,1]))~(1-pchoice_0[.,2])~pchoice_0[.,2] ;
umat2_a0 = zeros(numx,kparam) ;
umat2_a1 = (-(1-mstate[.,2]))~(1-pchoice_0[.,1])~pchoice_0[.,1] ;

zmat1 = (1-pchoice_0[.,1]).*umat1_a0 + pchoice_0[.,1].*umat1_a1 ;
zmat1 = inv_bf * zmat1 ;
zmat1 = (umat1_a1 + disfact * iptran1_a1 * zmat1)
- (umat1_a0 + disfact * iptran1_a0 * zmat1) ;

zmat2 = (1-pchoice_0[.,2]).*umat2_a0 + pchoice_0[.,2].*umat2_a1 ;
zmat2 = inv_bf * zmat2 ;
zmat2 = (umat2_a1 + disfact * iptran2_a1 * zmat2)
- (umat2_a0 + disfact * iptran2_a0 * zmat2) ;

emat1 = pdfn(cdfni(pchoice_0[.,1])) ;
emat1 = inv_bf * emat1 ;
emat1 = (disfact * iptran1_a1 * emat1)
- (disfact * iptran1_a0 * emat1) ;

emat2 = pdfn(cdfni(pchoice_0[.,2])) ;
emat2 = inv_bf * emat2 ;
emat2 = (disfact * iptran2_a1 * emat2)
- (disfact * iptran2_a0 * emat2) ;

zobs = zmat1[indobs,.] | zmat2[indobs,.] ;
eobs = emat1[indobs,.] | emat2[indobs,.] ;

// -----
// f. Pseudo Maximum Likelihood Estimation
// -----
{tetaest,varest,likelihood} = miprobit(yobs,zobs,eobs,zeros(kparam,1),0) ;
best[.,iter] = tetaest ;
varb[.,(iter-1)*kparam+1:iter*kparam] = varest ;
sdb      = sqrt(diag(varest)) ;
tstat   = tetaest./sdb ;

// -----
// g. Updating probabilities
// -----
pchoice_1 = cdfn(zmat1*tetaest +emat1)
~ cdfn(zmat2*tetaest +emat2) ;

// -----
// h. Checking for Convergence
// -----
if (iter==1) ;
criter = 1000 ;
elseif (iter>1) ;
criter = maxc(maxc(abs(best[.,iter]-best[.,iter-1]))) ;

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endif ;

// -----
// i. Presenting estimaton results from current NPL iteration
// -----

"" ;
"-----" ;
"      NPL ITERATION NUMBER      =" ; iter;
"      NPL Convergence Criterion   =" ; criter ;
"" ;
"      Estimation Results from Current Iteration";
"-----" ;
"      Log-Likelihood function = " likelihood ;
"" ;
"-----" ;
"      Parameter          Estimate      Standard      t-ratios";
"                  Errors" ;
"-----" ;
"" ;
k=1 ;
do while k<=kparam ;
  print $namesb[k];; tetaest[k];; sdb[k];; tstat[k];
  k=k+1 ;
endo;
"" ;
"-----" ;
"" ;

pchoice_0 = pchoice_1 ;
iter=iter+1 ;
endo ;

"" ;
"-----" ;
"-----" ;
"      NPL ESTIMATION" ;
"      CONVERGENCE ACHIEVED AFTER";; (iter-1);; "NPL ITERATIONS";
"-----" ;
"-----" ;

retp(best,varb,likelihood) ;
endp ;

// -----
// 11. NPL Estimation
// -----

"" ;
***** ;

```

```

"***** ;  

"    ESTIMATION OF THE STRUCTURAL PARAMETERS OF THE DYNAMIC GAME" ;  

"    BASED ON SIMULATED DATA FROM THE COMPUTED MPE" ;  

"***** ;  

"***** ;  

"" ;  

"-----" ;  

theta0 = zeros(kparam,1) ;  
  

{best1,varb,like1}  

= npldygam(aobs,aobs_1,freq_est_prob,dfact,vstate,theta0,npliter,namesb) ;  
  

"" ;  

"-----";  

"      True value of paramaters:" ;  

"" ;  

"          Entry Cost      =";; ecost ;  

"          Monopoly Profit =";; pi_mon ;  

"          Duopoly Profit   =";; pi_duo ;  

"" ;  

"-----";  

output off ;  
  

end ;

```