



Computational and Empirical Methods

Assignment VI

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%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
% Assignment 6 %
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clear          % Empty the Workspace
clc            % Clear the Screen
format compact % compact format

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%% 1 Cournot – Linear Case (15 points)

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%In a Cournot oligopoly, all firms produce and sell a homogeneous product. All firms'
output

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%is sold in a single market, for which there is a specified market demand. Each firm
chooses its

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%own quantity to maximize its own profits, but cannot directly influence the quantity
produced

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%by its rivals. Here, we consider a Cournot oligopoly with three firms that is
characterized by

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%the following inverse market demand function and individual firm cost function,
respectively:

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%P(Q) = 450 - 0.125Q

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%Ci(qi) = 50qi + 180

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%When each firm maximizes its own profit by selecting its own quantity, taking their
rivals'

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```

%quantities as given, we obtain the following First Order Conditions:

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%450 - 0.25q1 - 0.125q2 - 0.125q3 - 50 = 0

```

```

%450 - 0.125q1 - 0.25q2 - 0.125q3 - 50 = 0

```

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%450 - 0.125q1 - 0.125q2 - 0.25q3 - 50 = 0

```

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%% 1.1. Present the linear system of First Order Conditions in matrix form.

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%need to rewrite the equations as Aq = b

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% - 0.25q1 - 0.125q2 - 0.125q3 = -400

```

```

% - 0.125q1 - 0.25q2 - 0.125q3 = -400

```

```

% - 0.125q1 - 0.125q2 - 0.25q3 = -400

```

```

%[-0.25, -0.125, -0.125; -0.125, -0.25, -0.125; -0.125, -0.125, -0.25] *

```

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%[q1; q2; q3] = [-400; -400; -400]

```

```

%% 1.2. Solve it in Matlab via Matrix Inversion.

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```

A = [-0.25, -0.125, -0.125; -0.125, -0.25, -0.125; -0.125, -0.125, -0.25]; % set up A
matrix

```

```

b = [-400; -400; -400]; % set up b
matrix

```

```

q = inv(A)*b; % q = A^-1b

```

```

q

```

```

%q =

```

```

% 800

```

```

% 800

```

```

% 800

```

```

%% 1.3. Can you apply the iterative method for this problem that we covered in class?

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%Why or why not? If your answer is yes, find the three quantities via this approach.

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%Yes, with Gauss – Seidel Method.

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%Gauss–Jacobi Method (could not solve the problem since could not update with new

```

paramiters, as shown below)

```
q1 = 300;           % Initial guess
q2 = 300;
q3 = 300;

for i = 1:20
    q1_new = (- 0.125*q2 - 0.125*q3 + 400)/0.25; % Update q1
    q2_new = (- 0.125*q1 - 0.125*q3 + 400)/0.25; % Update q2
    q3_new = (- 0.125*q1 - 0.125*q2 + 400)/0.25; % Update q3
    q1 = q1_new;
    q2 = q2_new;
    q3 = q3_new;
    disp([q1, q2, q3]);
end
```

% 300; 300; 300

%Gauss-Jacobi Method, can solve the problem.

```
q1 = 300;           % Initial Guess
q2 = 300;
q3 = 300;

for i = 1:20
    q1 = (-0.125 * q2 - 0.125 * q3 + 400) / 0.25; % Update
    q2 = (-0.125 * q1 - 0.125 * q3 + 400) / 0.25;
    q3 = (-0.125 * q1 - 0.125 * q2 + 400) / 0.25;

    disp([q1, q2, q3]); % show each result
end
```

%800; 800; 800

%% 2 Cournot – Nonlinear Case (20 points)

%In class we have covered the Cournot example with 2 rms ($N = 2$) and $c = 2$.

% In this question, you'll compute numerically how the number of firms N affects equilibrium prices and quantities. Specifically, assume that there are N rms, each producing a quantity q_i . Total output in the economy is hence given by $Q = q_1 + q_2 + \dots + q_N$. Suppose the inverse demand for the good is given by

% $P(Q) = 1 - Q$

% and, as in class, each firm's total cost equals

% $TC = c \cdot (2/3) \cdot q^{(3/2)}$

% As before, assume that parameter $c = 2$.

%% 2.1. Find the equilibrium quantity and price in the case of a monopolist

% instead, i.e. in the case of 1 rm. Use fsolve().

% Profit = Total Revenue–Total Cost

% Profit = Price * Total Profuction –Total Cost

% Profit = $(1-q) \cdot q - 2 \cdot (2/3) \cdot q^{(3/2)}$

% first set up the function: HW6_Cournot

%calculation:

```
q0 = 1; % Pick an initial guess
Cournot_handle = @(q) HW6_Cournot(q); % Create the function handle = cricial
```

```

variable function(variable)
q_opt = fsolve(Cournot_handle, q0)           % Use fsolve (hundle, initial guess) to solve✓
the system

% Optimal production: Q_opt = 0.1340

p = 1 - q_opt
% eq price: p = 0.8660

%% 2.2. Find the equilibrium quantities and price in the case of 3 firms.
% P(Q) = 1 - q1 - q2 - q3
% TC = c*(2/3)*q^(3/2)
% Profit1 = (1 - q1 - q2 - q3)*q1 - 2*(2/3)*q1^(3/2)
% Profit2 = (1 - q1 - q2 - q3)*q2 - 2*(2/3)*q2^(3/2)
% Profit3 = (1 - q1 - q2 - q3)*q3 - 2*(2/3)*q3^(3/2)

% first set up the function: HW6_Cournot2

% calculation
q0 = [1;1;1];                               % Pick an initial guess
Cournot_handle = @(q) HW6_Cournot2(q);       % Create the function handle = cricial✓
variable function(variable)
q_opt = fsolve(Cournot_handle, q0)           % Use fsolve (hundle, initial guess) to solve✓
the system
%q_opt = [0.0955, 0.0955, 0.0955]

p = (1 - q_opt(1) - q_opt(2) - q_opt(3))
% p = 0.7135

%% 2.3. Based on your results, what do you think will be the equilibrium price when N ->✓
infinity
%when N -> infinity, the market will become perfectly competitive market.
%Then the market price will equal the marginal cost : P = 1 - Q = MC = 2 *
%q^(1/2)

%% 3 Finding the Equilibrium (30 points)
%Suppose you have an economy with a demand for cars of
% x = 2 - p
% supply
% x = -1 + exp(p)
%such that in equilibrium it must hold that
%2 - p = -1 + exp(p)
% <==> 3 - p - exp(p) = 0

%% 3.1. Given that equilibrium quantities cannot be negative,
% in what range will the equilibrium price need to be?

%Since x >= 0
% 1. 2 - p >= 0, 2. -1 + exp(p) >= 0
% 0 <= p <= 2

%% 3.2. Use this range and do a grid search to find the equilibrium price.
% Use increments of 0.01.

p = meshgrid(0:0.01:2);                     % build up the grid

for i = 1:size(p,1)                          % double loops

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for j = 1:size(p,2)
    diff(i,j) = 3 - p(i,j) - exp(p(i,j));    % function
    obj(i,j) = diff(i,j)^2;                  % objective is the diff^2
end
c_min = min(min(obj));                      % find the minimum of obj, the ideal✓
minimum should colse to 0
[i_max, j_max] = find(obj == c_min,1);
solution = p(i_max,j_max)
%solution = 0.7900

```

%% 3.3. Now instead, use Newton's method as covered in class to find the equilibrium price.✓

% Do not use fsolve(). Do you find the same price as you did with the grid search?
 $3 - p - \exp(p) = 0$

```

p = 6                                     % initial guess
for i = 1:20                             % iterate 20 times
    num = 3 - p - exp(p);                 % Numerator is the original geuition
    denom = -1 - exp(p);                  % Denominator is the FOC
    p = - num/denom + p;                  % The iteration starts
end

```

% p = 0.7921 the solution is different to what we got with grid search,
 % This discrepancy is likely due to us only using a second-order Taylor series,
 % which results in a loss of some accuracy.

%% 3.4. Now suppose you have two cars with demand functions

% $1x_1 + 0x_2 + 0.2p_1 - 0.3p_2 = 2$

% $0x_1 + 1x_2 - 0.3p_1 + 0.2p_2 = 2$

% and supply

% $1x_1 + 0x_2 - 1p_1 + 0p_2 = 1$

% $0x_1 + 1x_2 + 0p_1 - 1p_2 = 1$

%Set supply and demand equal for each type of car and solve for p1 and p2

% using matrix inversion.

%mertix:

A = [1,0,0.2,-0.3;0,1,-0.3,0.2;1,0,-1,0;0,1,0,-1];

b = [2;2;1;1];

x = A\b;

p1 = x(3)

p2 = x(4)

%solution: p1 = 1.1111, p2 = 1.1111.

%% 4 Pro t Maximization (35 points)

%Now, consider the problem of a firm that produces a good with production function

% $F(K; L) = 3K^{0.2} * L^{0.4}$,

% using two inputs: Capital at price 2 and labor at price 1. Formally, the firm

% problem is to maximize profits :

% $\max(K,L) = 3K^{0.2} * L^{0.4} - 2K - L$

%% 4.1. Solve this problem numerically using a grid search,

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% i.e. find the combination of K and L that maximizes profits pai.
% Assume that  $K \in [0,1]$  and  $L \in [0,1]$  and use increments of 0.01.
% FOC on K:  $0.6K^{(-0.8)} * L^{0.4} - 2 = 0$ 
% FOC on L:  $1.2K^{0.2} * L^{(-0.6)} - 1 = 0$ 
[K, L] = meshgrid(0:0.01:1,0:0.01:1); % set the grid

for i = 1:size(K,1) % set double loops
    for j = 1:size(K,2)
        pai1(i,j) = 0.6*K(i,j)^(-0.8)*L(i,j)^0.4 - 2; % profit function
        pai2(i,j) = 1.2*K(i,j)^0.2*L(i,j)^(-0.6) - 1;
        obj(i,j) = pai1(i,j)^2 + pai2(i,j)^2;
    end
end

c_min = min(min(obj)); % find the maximum of profit
[i_min,j_min] = find(obj == c_min);
solution = [K(i_min,j_min);L(i_min,j_min)]

% solution = 0.2000 0.8000

%% 4.2. Instead of using a grid search you can also derive the first-order conditions and solve those:
%  $0.6K^{(-0.8)} * L^{0.4} = 2$ 
%  $1.2K^{0.2} * L^{(-0.6)} = 1$ 
% Based on the discussions in class, what would be the most efficient way to solve these FOCs for K and L? Why?

%% 4.2.1. if we are asked to solve the maximization:
% Solution: Since we are doing maximization. Trust-Region Algorithm, which implies Newton's Method. Newton's Method has very good convergence properties and comparative fast. In addition, computing Gradient and Hessian manually and supplying it to the optimizer can substantially enhance the optimization.

%% 4.2.2. if we are asked to only solve this 2 equations:
% solution: The Newton's Method is the best, since the equations are not linear.

%% 4.3. Solve the system using the method you proposed in question (2).
%% 4.3.1. if we are asked to solve the maximization:
I0 = [1;1]; % Pick an initial guess
objective_handle = @(I) max4_2(I); % Create the function handle
options = optimoptions('fminunc','Algorithm','trust-region','GradObj','on','Hessian','on');
fminunc(objective_handle, I0, options)
% ans = 0.1972; 0.7887

%% 4.3.2. if we are asked to only solve this 2 equations:
%  $0.6K^{(-0.8)} * L^{0.4} = 2$ 
%  $1.2K^{0.2} * L^{(-0.6)} = 1$ 

% Pick an initial guess
K = 0.2;
L = 0.8;
I = [K; L];

for i = 1:10

```

```

F = [0.6*K^(-0.8)*L^0.4-2; 1.2*K^0.2*L^(-0.6)-1]; % Function
DF = [-0.48 * K^(-1.8) * L^(0.4), 0.24 * K^(-0.8) * L^(-0.6); 0.24 * K^(-0.8) * L^(-0.6),
-0.72 * K^(0.2) * L^(-1.6)]; % Gradient
I = - inv(DF) * F + I; % Newton
Step
K = I(1);
L = I(2);
disp(I');
end

% solution: K = 0.1972; L = 0.7887.

%% 4.4. Now instead suppose that the firm problem is
% max(K;L) pai(K,L) = 3K^(0.2)*L^(0.4)-e^(K+L)

%FOC on K = 0.2*3*K^(-0.8)*L^(0.4) - e^(K+L)
%FOC on L = 0.4*3*K^(0.2)*L^(-0.6) - e^(K+L)

I0 = [1;1]; % Pick an initial guess
max4_4_handle = @(I) max4_4(I); % Create the function handle
solution_opt = fsolve(max4_4_handle, I0) % Use fsolve to solve the system

%solution_opt = 0.1636; 0.3271

%% 5. Now instead, linearize the first-order conditions you obtained in (4)
% around K0 = 1 and L0 = 2 and solve this linearized system using matrix
% inversion. Is the solution close to the one you found in (4)?

%No.

% unliner FOC:
%FOC on K = 0.2*3*K^(-0.8)*L^(0.4) - e^(K+L)
%FOC on L = 0.4*3*K^(0.2)*L^(-0.6) - e^(K+L)

%Tylar Extension:
%FOC1_taylor = (0.6 * 2^(0.4) - exp(3)) + (-0.48 * 2^(0.4) - exp(3)) * (K - 1) + (0.24 *
2^(-0.6) - exp(3)) * (L - 2);
%FOC2_taylor = (1.2 * 2^(-0.6) - exp(3)) + (0.24 * 2^(-0.6) - exp(3)) * (K - 1) + (-0.72
* 2^(-1.6) - exp(3)) * (L - 2);

%rearrange:
% (-0.48 * 2^(0.4) - exp(3)) * K + (0.24 * 2^(-0.6) - exp(3)) * L = -(1.08 * 2^(0.4) -
0.48 * 2^(-0.6) + 2 * exp(3))
% (0.24 * 2^(-0.6) - exp(3)) * K + (-0.72 * 2^(-1.6) - exp(3)) * L = -(0.96 * 2^(-0.6) +
1.44 * 2^(-1.6) + 2 * exp(3))

% rearrange to be metrix:
A = [-0.48 * 2^(0.4) - exp(3), 0.24 * 2^(-0.6) - exp(3); 0.24 * 2^(-0.6) - exp(3), -0.72
* 2^(-1.6) - exp(3)];
b = [-(1.08 * 2^(0.4) - 0.48 * 2^(-0.6) + 2 * exp(3)); -(0.96 * 2^(-0.6) + 1.44 * 2^
(-1.6) + 2 * exp(3))];

X = A\b;
K = X(1)
L = X(2)

% solution: K = 0.6815; L = 1.3630. No, they are not close.

```



```
function pai_der1 = HW6_Cournot(q)
pai_der1 = 1 - 2 * q - 2*q^(1/2)    % calculate the derivative of pai
end
```

```
function pai_der1 = HW6_Cournot2(q)
q1 = q(1)
q2 = q(2)
q3 = q(3)

pai1_der1 = 1 - 2 * q1 - q2 - q3 - 2 * q1^(1/2); % calculate the derivative of pai1
pai2_der1 = 1 - q1 - 2 * q2 - q3 - 2 * q1^(1/2); % calculate the derivative of pai2
pai3_der1 = 1 - q1 - q2 - 2 * q3 - 2 * q1^(1/2); % calculate the derivative of pai3

pai_der1 = [pai1_der1; pai2_der1; pai3_der1]; % return
end
```

```

function [f,g,H] = max4_2(I)
K = I(1);
L = I(2);

f = 3 * K^(0.2) * L^(0.4) - 2*K - L;
f = -f; % since max

% Supply the Gradient
%-----
if nargout > 1 % nargout: Number of output arguments
    g = [0.6*K^(-0.8)*L^(0.4)-2;
         1.2*K^(0.2)*L^(-0.6)-1];
    g = -g; % since max
end

% Supply the Hessian
%-----
if nargout > 2
    H = [-0.8*0.6*K^(-1.8)*L^0.4, 0.6*0.4*K^(-0.8)*L^(-0.6); 0.2*1.2*K^(-0.8)*L^(-0.6), -1.2*0.6*K^0.2*L^(-1.6)];
    H = -H; % since max
end

end

```

```
function F = max4_4(I)

K = I(1);           % Define K as first element of I
L = I(2);           % Define L as second element of I

F = [0.2*3*K^(-0.8)*L^(0.4) - exp(K+L); 0.4*3*K^(0.2)*L^(-0.6) - exp(K+L)]
                                % Compute the function values for the given K,L
disp([K,L])
end
```