

Advanced Macro Theory  
ECON 5300, University of Oslo:

A Matlab Primer

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# I. Why Matlab?

- Matlab is optimized for matrix computations that are often used in many fields of economics.
- Many other applied economists use Matlab to solve and simulate numerical models. You can learn and profit from other people's codes.
- Matlab is a high-level programming language that does the computational housekeeping for you.
- Matlab is a proprietary software and well documented. Advanced users can also use Octave the open source clone of Matlab.
- The University of Oslo has a campus license for Matlab that you can access through the UiO Program Kiosk

<http://www.uio.no/english/services/it/computer/software/servers>

But the best option is to get it installed on your own computer (consult the IT help desk!).

## II. The Basics

### A. *Some Basic Commands*

- **help:**  
Lists all toolboxes and subfolders that are on your search path. With the command `help cmd` you get access to the help file of the specific command `cmd`. A more user-friendly documentation can be accessed by typing `helpbrowser` in the command line.
- **edit:**  
Opens a new script file with `.m` extension in the text editor. Save all your code in scripts. Type `edit fun` to open the existing file `fun.m` in the text editor.
- **%:**  
It is very important to document your code. Put a `%` and Matlab will ignore the rest of the command line. This leaves space for comments and explanations.

## B. *Some Matrix Commands*

- $A = [1, 3; 2, 4]$ ; initializes the 2 by 2 matrix

$$A = \begin{bmatrix} 1 & 3 \\ 2 & 4 \end{bmatrix}.$$

- Let's look at some matrix operators,  $+$ ,  $-$ ,  $*$ ,  $.*$ ,  $./$ ,  $\wedge$ , where a dot indicates an element-wise operation. Take the previous matrix  $A$ , then

$$B \equiv A * A = \begin{bmatrix} 7 & 15 \\ 10 & 22 \end{bmatrix} \quad \text{and} \quad C \equiv A .* A = \begin{bmatrix} 1 & 9 \\ 4 & 16 \end{bmatrix}$$

- $D = A(1, :)$ ; returns the first row of matrix  $A$ ,  $[1 \ 3]$ .
- $E = A(:, 2)$ ; returns the second column of matrix  $A$ ,  $[3 \ 4]'$ .
- $i = 2:2:10$ ; creates the row vector  $i = [2 \ 4 \ 6 \ 8 \ 10]$ .
- $F = \max(A)$ ; returns the max along the columns of  $A$ ,  $[2 \ 4]$ .
- $G = \max(A, [], 2)$ ; returns the max along the rows of  $A$ ,  $[3 \ 4]'$ .

## C. *Conditional Statements and Looping*

```
%% draws random grades for the macro students

% cell of strings with students' names
students = {'Frikk''s', 'Federica''s', 'Yudi''s', 'Ola''s'};
n = size(students,2); % class size

for i=1:n % loop over n students
    student = students{i}; % reference ith student
    points = randi(100,1,1); % random integer between 1 and 100
    if points>50
        grade = 6; % max grade in Switzerland
    else
        grade = 4; % pass grade in Switzerland
    end
    % print grade for ith student
    fprintf('%-10s grade is a %i.\n',student,grade)
end
```

## D. *Scripts and Functions*

```
%% allocate deterministic grades for the macro students

% cell of strings with student names
students = {'Frikk''s', 'Federica''s', 'Yudi''s', 'Ola''s'};
n = size(students,2); % class size

for i=1:n % loop over n students
    student = students{i};          % reference ith student
    grade   = get_grade(student);  % calls function get_grade()
    % print results
    fprintf('%-10s grade is a %8.6f.\n',student,grade)
end

% function get_grade() is located in a separate .m file in the
% same folder or on the search path.
```

```
%% returns deterministic grades for the macro students

function grade = get_grade(student)

% search for student name that matches and allocate grade
switch student

    case 'Frikk''s'
        grade = 5.999999;
    case 'Federica''s'
        grade = 5.999998;
    case 'Yudi''s'
        grade = 5.999997;
    case 'Ola''s'
        grade = 5.999996;
    end

end
```

## E. *Programming Style*

- Document your code, choose meaningful names for your variables, be transparent.
- “Get it run right, then get it run fast.”
- Avoid loops whenever possible. In Matlab loops are generally slower than direct matrix operations. The code

```
n = 106;  
for i=1:n, y(i)=log(i); end
```

is on my PC around 7 times slower than

```
y = log(1:1:106);
```

The problem is squared if you work with nested loops. Use `tic` before and `toc` after a statement to measure running times.



## F. *Rootfinding and Anonymous Functions*

- Consider the steady-state condition for the capital stock in the Solow Model

$$sf(k^*) = \delta k^*.$$

- Solving for  $k^*$  is the same as finding the root (the zero) of

$$h(k) \equiv sf(k) - \delta k.$$

- Solution Algorithm:
  1. Set parameters.
  2. Define anonymous functions (not located in a separate file).
  3. Use the Matlab built-in routine `fsolve` to find the root of  $h(k)$ .

```

%% solves for the steady-state capital stock in the Solow Model

% parameters
alpha = .33;    % capital income share
delta = .10;    % depreciation rate
s      = .30;    % saving rate

% define anonymous functions
f = @(k) k.^alpha;
h = @(k) s*f(k)-delta*k;

% solve for the root of h(k)
k0 = 5;          % initial guess
options = optimset('Display','iter'); % display iterations
kstar = fsolve(h,k0,options);

```

## G. *Further Literature*

- Miranda, Mario J., and Paul L. Fackler (2002). “Applied Computational Economics and Finance,” MIT Press. Their book is accompanied by the CompEcon Toolbox for Matlab

<http://www4.ncsu.edu/~pfackler/compecon>

- Adda, Jérôme, and Russell Cooper, “Dynamic Economics,” MIT Press.
- Sigmon, Kermit (1993). “Matlab Primer (Third Edition),” University of Florida.
- Official website, [www.mathworks.com](http://www.mathworks.com).
- Judd, Kenneth (1998). “Numerical Methodes in Economics,” MIT Press.

### III. An Example

#### A. Neoclassical Growth Model

- Bellman equation

$$V(k) = \max_{0 \leq k' \leq f(k) + (1-\delta)k} u(c(k, k')) + \beta V(k'), \quad c(k, k') = f(k) + (1 - \delta)k - k'.$$

- Equilibrium conditions are a system of three functional equations in  $V(k)$ ,  $k' = g(k)$  and  $c(k)$

$$\begin{aligned} V(k) &= u(c(k)) + \beta V(g(k)) \\ \frac{u_1(c(k))}{u_1(c(g(k)))} &= \beta [f_1(g(k)) + (1 - \delta)] \\ c(k) &= f(k) + (1 - \delta)k - g(k), \end{aligned}$$

where  $k$  is the state variable while  $u(c)$  and  $f(k)$  are known functions to be specified.

## B. *Solution Algorithm*

1. Choose functional forms and parameters of preferences,  $u(c)$ , and technology,  $f(k)$ .
2. Compute the steady-state capital stock,  $k^* = g(k^*)$ .
3. Discretize the state space,  $k \in (0, \infty)$ , on the subgrid  $[k_{min}, k_{max}]$  around the steady-state,  $k^*$ .
4. Iterate on the value function,

$$V^{j+1}(k) = \max_{k' \in [k_{min}, f(k) + (1-\delta)k]} u(c(k, k')) + \beta V^j(k'),$$

until convergence, starting with a guess  $V^0(k')$ .

5. Plot  $V(k)$ ,  $g(k)$  and  $c(k)$  over  $k$ .

```

%% 1. functional forms and parameters

% parameters
alpha = .36;      % capital income share
beta  = .95;      % subjective discount factor
delta = .10;      % depreciation rate
sigma = 2;        % relative risk aversion parameter

% functional forms
u = @(c) c.^(1-sigma)/(1-sigma); % preferences
f = @(k) k.^alpha; % technology
f1 = @(k) alpha*k.^(alpha-1); % rental rate

```

```
%% 2. solve for steady-state capital stock

% use anonymous function
steady_state_condition = @(k) beta*(f1(k)+(1-delta))-1;

k0 = 4; % initial guess
options = optimset('Display','iter'); % display iterations
% use built-in rootfinding
kstar = fsolve(steady_state_condition,k0,options);
```

```
%% 3. state space discretization

nk = 500;          % number of grid points

% set up grid around the steady state
range = .10;
kmin  = (1-range)*kstar;    % lower bound of the grid
kmax  = (1+range)*kstar;    % upper bound of the grid

% equally spaced grid
kgrid = linspace(kmin,kmax,nk)';

% transformation into matrices that can be used for the evaluation
% of functions. kmat varies along columns, kpmat varies along rows.
[kmat,kpmat] = ndgrid(kgrid,kgrid);    % kp denotes k'
```



Let's look into the matrices:

- $k_{\text{grid}} = [k_1 k_2 \dots k_{n_k}]'$  is the discretized state  $k$  with  $k_1 = k_{\text{min}}$ ,  $k_{n_k} = k_{\text{max}}$  and  $n_k$  grid points.
- $k_{\text{mat}}$  is the  $n_k$  by  $n_k$  matrix (grid points vary along columns,  $k$ )

$$\begin{bmatrix} k_1 & \dots & k_1 \\ \vdots & \ddots & \vdots \\ k_{n_k} & \dots & k_{n_k} \end{bmatrix}.$$

- $k_{\text{pmat}}$  is the  $n_k$  by  $n_k$  matrix (grid points vary along rows,  $k'$ )

$$\begin{bmatrix} k_1 & \dots & k_{n_k} \\ \vdots & \ddots & \vdots \\ k_1 & \dots & k_{n_k} \end{bmatrix}.$$

- $c(k_{\text{mat}}, k_{\text{pmat}})$  then delivers the consumption level for any combination of  $k$  and  $k'$  on  $k_{\text{grid}}$ .

```

%% 4. value function iteration

% initial guess: one-period problem, thus kp=0.
V = u(f(kgrid)+(1-delta)*kgrid);

% continuation values and consumption possibilities using any
% combination of k and kp.
Vmat = u(f(kpmat)+(1-delta)*kpmat);
cmat = f(kmat)+(1-delta)*kmat-kpmat;

% momentary utility
% map negative consumption values into very low utility level.
umat = (cmat<=0).*(-10^6)+(cmat>0).*u(cmat);

% set convergence tolerance
tol = 10^(-8);
maxit = 500;
fprintf('iter \t norm \n');

```

```

tic
for j=1:maxit
    % search for kp = argmax umat+beta*Vmat for each k
    % max is along dimension two (row)
    [Vnext,indices] = max(umat+beta*Vmat, [],2);
    error = norm(Vnext-V);
    fprintf('%4i %2.1e \n',j,error);
    % stopping rule
    if error < tol
        fprintf('Elapsed Time = %4.2f Seconds\n',toc);
        break;
    else
        V = Vnext;
        Vmat = repmat(V',nk,1);
    end
end

k = kgrid; kp = kgrid(indices); c = f(k)+(1-delta)*k-kp;

```

Let's look into the matrices (I chose indices arbitrarily!):

- Grid search for the maximum returns:

$k$	$V^{j+1}(k)$	indices <sup><math>j+1</math></sup>	$k' = g^{j+1}(k)$	$V^{j+1}(k')$
$k_1$	$V^{j+1}(k_1)$	2	$k_2$	$V^{j+1}(k_2)$
$k_2$	$V^{j+1}(k_2)$	3	$k_3$	$V^{j+1}(k_3)$
$\vdots$	$\vdots$	$\vdots$	$\vdots$	
$k_{n_k}$	$V^{j+1}(k_{n_k})$	$n_k - 1$	$k_{n_k-1}$	$V^{j+1}(k_{n_k-1})$

- Value function matrix for the next iteration only varies with  $k'$  but not with  $k$ .

$$\text{Vmat}^{j+1} = \begin{bmatrix} V^{j+1}(k_1) & \dots & V^{j+1}(k_{n_k}) \\ \vdots & \ddots & \vdots \\ V^{j+1}(k_1) & \dots & V^{j+1}(k_{n_k}) \end{bmatrix} = \begin{bmatrix} V^{j+1}(k)' \\ \vdots \\ V^{j+1}(k)' \end{bmatrix}.$$

```
%% 5. plot unknown functions
```

```
scrsz = get(0,'ScreenSize');  
figure('Position',[scrsz(3)*1/4 scrsz(4)*1/4 scrsz(3)*1/2 ...  
    scrsz(4)*1/2]);
```

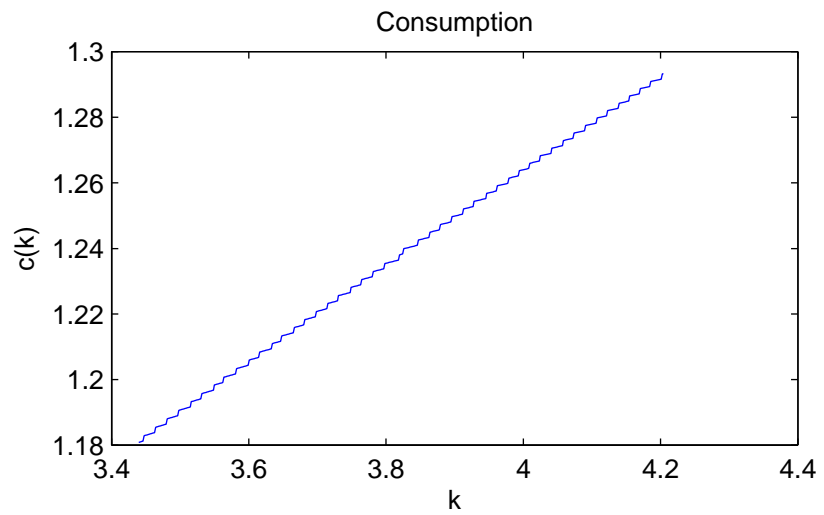
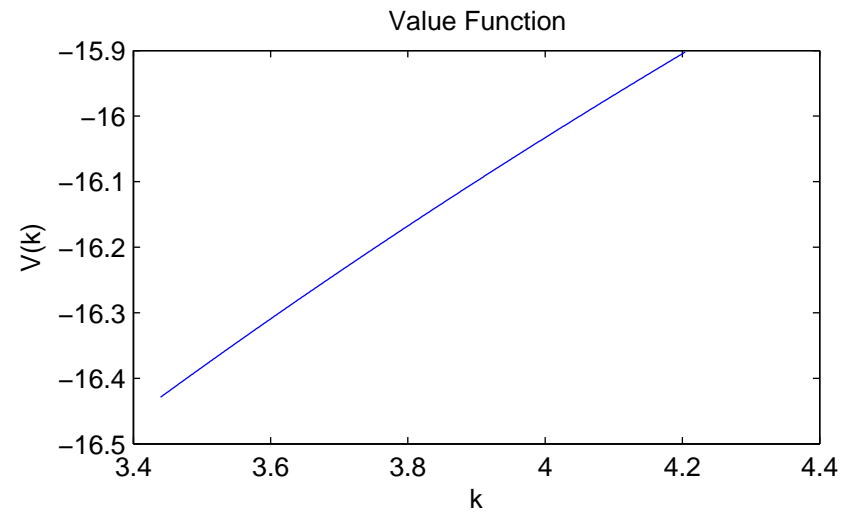
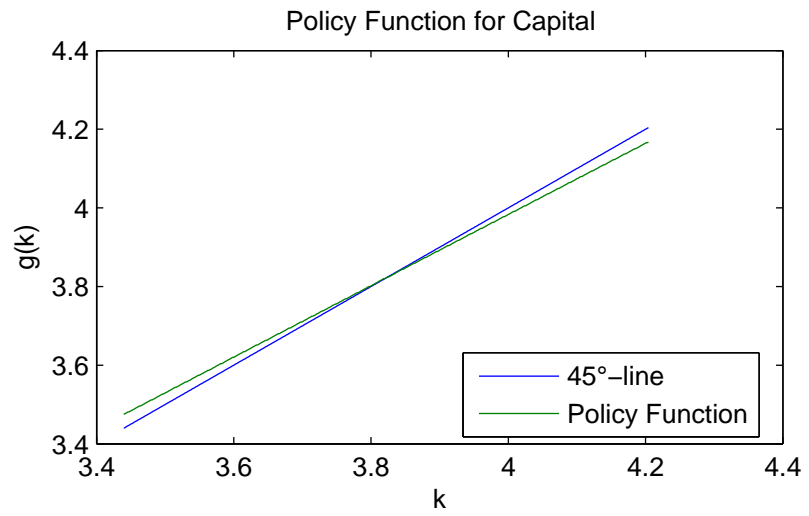
```
% policy function
```

```
subplot(2,2,1)  
plot(k,[k,kp]);          % plot k and kp at the same time  
title('Policy Function for Capital');  
xlabel('k'); ylabel('g(k)');  
legend('45°-line','Policy Function','Location','Best');
```

```
% value function
```

```
subplot(2,2,2)  
plot(k,V);  
title('Value Function');  
xlabel('k'); ylabel('V(k)');
```

```
% consumption
subplot(2,2,3)
plot(k,c);
title('Consumption');
xlabel('k'); ylabel('c(k)');
```



## IV. Exercise 3.3: Howard's Policy Iteration

- The most time consuming part in the grid search algorithm of Section III is to find the policy function  $g^j(k)$  for each state  $k$  in each iteration  $j$ .
- You can speed up the algorithm by iterating on the policy function  $g^j(k)$  to update the value function  $V^j(k)$  many times

$$V^{j,h+1}(k) = u(c^j(k)) + \beta V^{j,h}(g^j(k)), \quad c^j(k) = f(k) + (1 - \delta)k - g^j(k),$$

before you continue to update the policy function.

- Exercise: implement Howard's policy iteration in the algorithm of Section III.



## V. Summary

- Matlab provides powerful tools to solve dynamic economic models.
- Learn from other people's codes.
- Write code that other people can learn from.