

# Advanced Macroeconomics

Lecture 5: growth theory  
and dynamic optimization, part four

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1st Semester 2019

# This class

- Analyzing the dynamics of the Ramsey-Cass-Koopmans model
  - a system of nonlinear difference equations
  - log-linearization (convenient local approximation)
  - solving model by method of undetermined coefficients
  - examples and introduction to Matlab

# Log-linearization

- Consider scalar function

$$y_t = f(x_t)$$

with steady state satisfying  $\bar{y} = f(\bar{x})$ . Local deviation in levels

$$y_t - \bar{y} \approx f'(\bar{x})(x_t - \bar{x})$$

or

$$\frac{y_t - \bar{y}}{\bar{y}} \approx \left( \frac{f'(\bar{x})\bar{x}}{f(\bar{x})} \right) \frac{x_t - \bar{x}}{\bar{x}}$$

- Note the following approximation for log-deviations

$$\hat{y}_t \equiv \log \left( \frac{y_t}{\bar{y}} \right) = \log \left( 1 + \frac{y_t - \bar{y}}{\bar{y}} \right) \approx \frac{y_t - \bar{y}}{\bar{y}}$$

# Log-linearization

- Hence in log-deviations coefficients are elasticities (units free)

$$\hat{y}_t \approx \left( \frac{f'(\bar{x})\bar{x}}{f(\bar{x})} \right) \hat{x}_t$$

- Generalizes naturally to multivariate functions, if  $y_t = f(x_t, z_t)$  with  $\bar{y} = f(\bar{x}, \bar{z})$  then

$$\hat{y}_t \approx \frac{f_x(\bar{x}, \bar{z})\bar{x}}{f(\bar{x}, \bar{z})} \hat{x}_t + \frac{f_z(\bar{x}, \bar{z})\bar{z}}{f(\bar{x}, \bar{z})} \hat{z}_t$$

where  $f_x$  and  $f_z$  denote partial derivatives

# Examples

- Power functions

$$y = x^a z^b$$

implies

$$\hat{y} = a\hat{x} + b\hat{z}$$

(of course, since exactly log-linear)

- Linear functions

$$y = ax + bz$$

implies

$$\hat{y} = \left( \frac{a\bar{x}}{a\bar{x} + b\bar{z}} \right) \hat{x} + \left( \frac{b\bar{z}}{a\bar{x} + b\bar{z}} \right) \hat{z}$$

# Examples

- Combine these rules to log-linearize more complex expressions

$$y = f(x) = (ax + b)^c$$

- Gives

$$\hat{y} = c \left( \frac{a\bar{x}}{a\bar{x} + b} \right) \hat{x}$$

# Ramsey-Cass-Koopmans growth model

- Recall dynamical system in consumption  $c_t$  and capital  $k_t$

$$u'(c_t) = \beta u'(c_{t+1}) [f'(k_{t+1}) + 1 - \delta] \quad (\text{Euler eq.})$$

and

$$c_t + k_{t+1} = f(k_t) + (1 - \delta)k_t \quad (\text{resource constraint})$$

with given initial condition and transversality condition

- Next steps: (i) log-linearization around steady state  $\bar{c}, \bar{k}$  then  
(ii) determine magnitudes of eigenvalues, stability

# Log-linearizing the growth model

- Resource constraint in levels

$$c_t + k_{t+1} = f(k_t) + (1 - \delta)k_t$$

- Log-linearized version, treated as exact

$$\bar{c}\hat{c}_t + \bar{k}\hat{k}_{t+1} = f'(\bar{k})\bar{k}\hat{k}_t + (1 - \delta)\bar{k}\hat{k}_t$$

- Recall that in steady state

$$1 = \beta[f'(\bar{k}) + 1 - \delta]$$

- Using this to simplify, gives

$$\bar{c}\hat{c}_t + \bar{k}\hat{k}_{t+1} = \frac{1}{\beta}\bar{k}\hat{k}_t$$



# Log-linearizing the growth model

- Consumption Euler equation in levels

$$u'(c_t) = \beta u'(c_{t+1}) R_{t+1}, \quad R_{t+1} \equiv f'(k_{t+1}) + 1 - \delta$$

- Log-linearized version, treated as exact

$$u''(\bar{c}) \bar{c} \hat{c}_t = \beta u''(\bar{c}) \bar{R} \bar{c} \hat{c}_{t+1} + \beta u'(\bar{c}) \bar{R} \hat{R}_{t+1}$$

where

$$\bar{R} \hat{R}_{t+1} = f''(\bar{k}) \bar{k} \hat{k}_{t+1} \quad \text{and} \quad \beta \bar{R} = 1$$

- Notation: the Arrow/Pratt measure of *relative risk aversion*

$$\sigma(c) \equiv -\frac{u''(c)c}{u'(c)} > 0$$

# Log-linearizing the growth model

- With this notation

$$\hat{c}_t = \hat{c}_{t+1} - \frac{1}{\sigma(\bar{c})} \hat{R}_{t+1}$$

or

$$\hat{c}_{t+1} - \hat{c}_t = \frac{1}{\sigma(\bar{c})} \hat{R}_{t+1}, \quad \hat{R}_{t+1} = \beta f''(\bar{k}) \bar{k} \hat{k}_{t+1}$$

- Note consumption growing,  $\hat{c}_{t+1} > \hat{c}_t$ , when return on capital is relatively high,  $\hat{R}_{t+1} > 0$ , i.e., when capital stock will be below steady state,  $\hat{k}_{t+1} < 0$  (just as in the phase diagram)
- The coefficient  $\frac{1}{\sigma(\bar{c})}$  is a measure of the *intertemporal elasticity of substitution* (willingness to substitute consumption over time)

# Log-linearized growth model

- In short, pair of equations

$$\hat{c}_{t+1} = \hat{c}_t + \frac{\beta f''(\bar{k})\bar{k}}{\sigma(\bar{c})} \hat{k}_{t+1}$$

and

$$\bar{c}\hat{c}_t + \bar{k}\hat{k}_{t+1} = \frac{1}{\beta} \bar{k}\hat{k}_t$$

- Implies a system of difference equations

$$\begin{pmatrix} \hat{c}_{t+1} \\ \hat{k}_{t+1} \end{pmatrix} = \begin{pmatrix} 1 - \frac{\beta f''(\bar{k})\bar{c}}{\sigma(\bar{c})} & \frac{f''(\bar{k})\bar{k}}{\sigma(\bar{c})} \\ -\frac{\bar{c}}{\bar{k}} & \frac{1}{\beta} \end{pmatrix} \begin{pmatrix} \hat{c}_t \\ \hat{k}_t \end{pmatrix}$$

What are the eigenvalues of this system?

# Eigenvalues of the growth model

- Recall that trace of coefficient matrix  $\mathbf{A}$  is sum of eigenvalues

$$\text{tr}(\mathbf{A}) = \lambda_1 + \lambda_2$$

and determinant of  $\mathbf{A}$  is product of eigenvalues

$$\det(\mathbf{A}) = \lambda_1 \lambda_2$$

- Gives

$$\lambda_1 + \lambda_2 = 1 - \frac{\beta f''(\bar{k})\bar{c}}{\sigma(\bar{c})} + \frac{1}{\beta} > 2$$

and

$$\lambda_1 \lambda_2 = \frac{1}{\beta} > 1$$

- Hence both roots positive and at least one is explosive

# Eigenvalues of the growth model

- Recall characteristic polynomial can be written

$$p(\lambda) = \lambda^2 - \text{tr}(\mathbf{A})\lambda + \det(\mathbf{A}) = (\lambda - \lambda_1)(\lambda - \lambda_2)$$

- Now consider polynomial  $p(\lambda)$  evaluated at  $\lambda = 1$

$$p(1) = (1 - \lambda_1)(1 - \lambda_2) > 0 \quad \Leftrightarrow \quad \text{both roots on same side of } +1$$

- For the growth model we have

$$p(1) = 1 - \left(1 - \frac{\beta f''(\bar{k})\bar{c}}{\sigma(\bar{c})} + \frac{1}{\beta}\right) + \frac{1}{\beta} = \frac{\beta f''(\bar{k})\bar{c}}{\sigma(\bar{c})} < 0$$

- Hence one stable root  $0 < \lambda_1 < 1$ , one explosive root  $\lambda_2 > 1/\beta > 1$

# Implications for stability

- As we saw in the phase diagram, system is *saddle path* unstable
- For almost all initial conditions, system diverges from steady state
- Initial capital  $\hat{k}_0$  pre-determined, given exogenously
- Initial consumption  $\hat{c}_0$  *not* pre-determined, jumps to *stable arm*
- Compute jump by looking at the solution for the linearized system

# Implications for stability

- Recall

$$\begin{pmatrix} \hat{c}_{t+1} \\ \hat{k}_{t+1} \end{pmatrix} = \mathbf{A} \begin{pmatrix} \hat{c}_t \\ \hat{k}_t \end{pmatrix}$$

hence using  $\mathbf{A} = \mathbf{V} \mathbf{\Lambda} \mathbf{V}^{-1}$  and iterating forward

$$\begin{pmatrix} \hat{c}_t \\ \hat{k}_t \end{pmatrix} = \mathbf{V} \mathbf{\Lambda}^t \mathbf{V}^{-1} \begin{pmatrix} \hat{c}_0 \\ \hat{k}_0 \end{pmatrix}$$

- Writing this out explicitly

$$\hat{c}_t = v_{11} \frac{v_{22}\hat{c}_0 - v_{12}\hat{k}_0}{v_{11}v_{22} - v_{12}v_{21}} \lambda_1^t - v_{12} \frac{v_{21}\hat{c}_0 - v_{11}\hat{k}_0}{v_{11}v_{22} - v_{12}v_{21}} \lambda_2^t$$

and

$$\hat{k}_t = v_{21} \frac{v_{22}\hat{c}_0 - v_{12}\hat{k}_0}{v_{11}v_{22} - v_{12}v_{21}} \lambda_1^t - v_{22} \frac{v_{21}\hat{c}_0 - v_{11}\hat{k}_0}{v_{11}v_{22} - v_{12}v_{21}} \lambda_2^t$$

## Initial jump $\hat{c}_0$

- If  $\lambda_2$  is the unstable root, then setting

$$\hat{c}_0 = \frac{v_{11}}{v_{21}} \hat{k}_0$$

will neutralize the explosive dynamics

- Plugging this jump for initial consumption gives

$$\hat{c}_t = v_{11} \frac{v_{22} \frac{v_{11}}{v_{21}} - v_{12}}{v_{11}v_{22} - v_{12}v_{21}} \lambda_1^t \hat{k}_0 = \frac{v_{11}}{v_{21}} \lambda_1^t \hat{k}_0$$

and similarly for capital accumulation

$$\hat{k}_t = v_{21} \frac{v_{22} \frac{v_{11}}{v_{21}} - v_{12}}{v_{11}v_{22} - v_{12}v_{21}} \lambda_1^t \hat{k}_0 = \lambda_1^t \hat{k}_0$$



# Simple representation

- Let  $\lambda \equiv \lambda_1$  denote the stable root
- Solution for capital accumulation

$$\hat{k}_{t+1} = \lambda \hat{k}_t, \quad t = 0, 1, \dots, \quad \hat{k}_0 \text{ given}$$

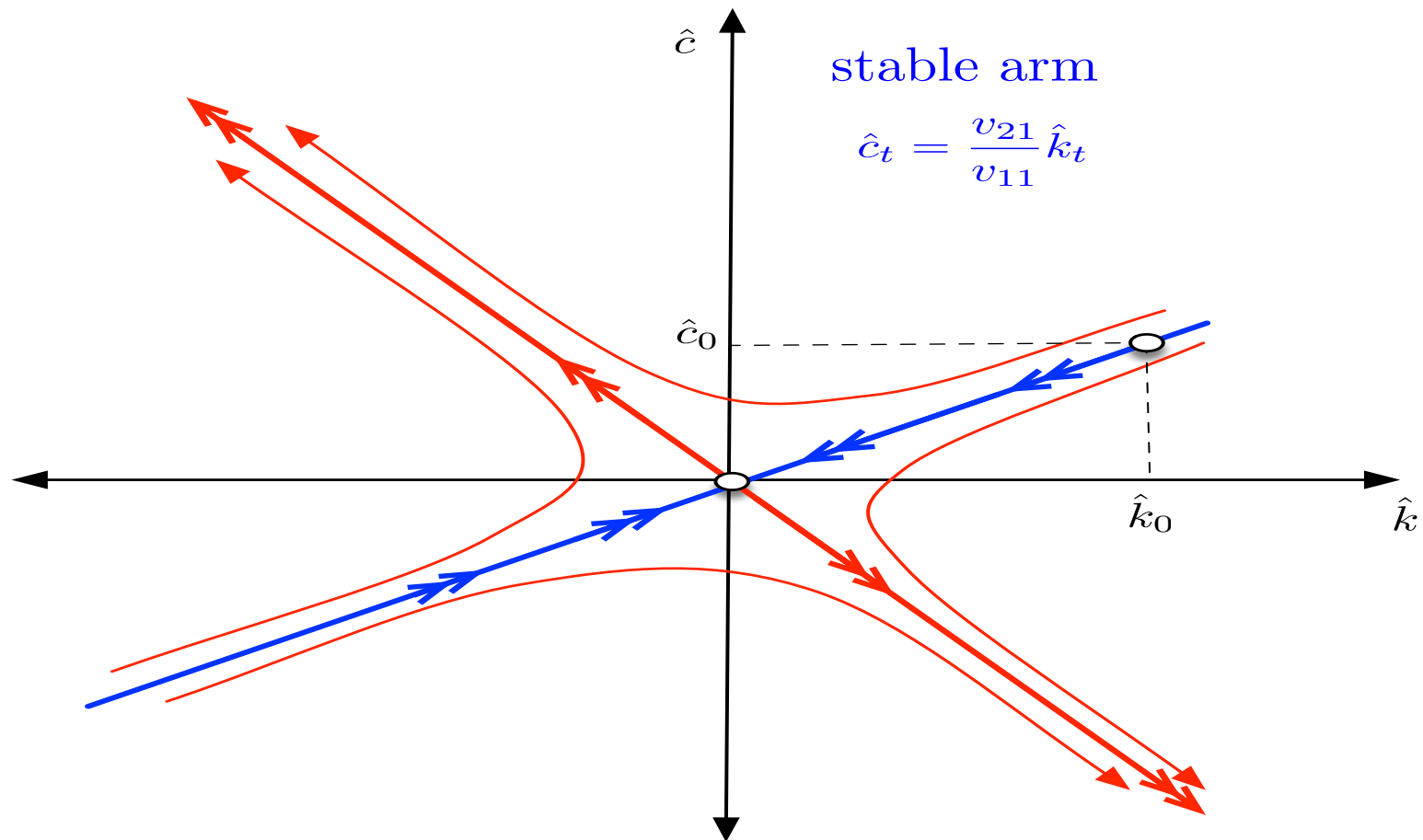
- Solution for consumption, inherits dynamics of capital

$$\hat{c}_t = \frac{v_{11}}{v_{21}} \hat{k}_t, \quad t = 0, 1, \dots$$

This function *is* the stable arm of the saddle path

- Both capital and consumption converge monotonically to steady-state, faster convergence the smaller is  $|\lambda|$

# Saddle path in log-deviations



# Method of undetermined coefficients

- Direct approach that will be useful throughout course
- Guess linear law of motion for capital

$$\hat{k}_{t+1} = \psi_{kk} \hat{k}_t$$

and likewise for consumption

$$\hat{c}_t = \psi_{ck} \hat{k}_t$$

- Two coefficients that have to be determined

$$\psi_{kk} , \psi_{ck}$$

# Method of undetermined coefficients

- Resource constraint

$$\bar{c}\hat{c}_t + \bar{k}\hat{k}_{t+1} = \frac{1}{\beta}\bar{k}\hat{k}_t$$

- Plug in guesses and collect terms

$$\left[ \bar{c}\psi_{ck} + \bar{k}\psi_{kk} - \frac{1}{\beta}\bar{k} \right] \hat{k}_t = 0$$

- This has to hold for *any* value of  $\hat{k}_t$ . Gives condition

$$\boxed{\bar{c}\psi_{ck} + \bar{k}\psi_{kk} - \frac{1}{\beta}\bar{k} = 0}$$

# Method of undetermined coefficients

- Consumption Euler equation

$$\hat{c}_{t+1} = \hat{c}_t + \frac{\beta f''(\bar{k})\bar{k}}{\sigma(\bar{c})} \hat{k}_{t+1}$$

- Plug in guesses and collect terms

$$\left[ \psi_{ck} \psi_{kk} - \psi_{ck} - \frac{\beta f''(\bar{k})\bar{k}}{\sigma(\bar{c})} \psi_{kk} \right] \hat{k}_t = 0$$

- This has to hold for *any* value of  $\hat{k}_t$ . Gives condition

$$\boxed{\psi_{ck} \psi_{kk} - \psi_{ck} - \frac{\beta f''(\bar{k})\bar{k}}{\sigma(\bar{c})} \psi_{kk} = 0}$$

- Two equations in two unknowns  $\psi_{ck}, \psi_{kk}$

# Method of undetermined coefficients

- Use first condition to solve for  $\psi_{ck}$  in terms of  $\psi_{kk}$

$$\psi_{ck} = \left( \frac{1}{\beta} - \psi_{kk} \right) \frac{\bar{k}}{\bar{c}}$$

- Then plug this into second condition and rearrange to get a *quadratic* in  $\psi_{kk}$

$$\psi_{kk}^2 - \left( 1 - \frac{\beta f''(\bar{k})\bar{c}}{\sigma(\bar{c})} + \frac{1}{\beta} \right) \psi_{kk} + \frac{1}{\beta} = 0$$

- Looks familiar. This is *exactly the same* characteristic polynomial we had before. Two solutions, one stable and one unstable

# Solution

- In short, the coefficient  $\psi_{kk}$  is the stable eigenvalue

$$\psi_{kk} = \lambda \in (0, 1)$$

Can then recover the slope of the stable-arm

$$\psi_{ck} = \left( \frac{1}{\beta} - \psi_{kk} \right) \frac{\bar{k}}{\bar{c}} > 0$$

- So far, have simply confirmed the intuition in the phase diagram
- But want to go further than that, want to actually solve the model
- To do this, need specific functional forms and parameter values
- Can then calculate  $\psi_{kk}$ ,  $\psi_{ck}$  and study dynamics



# Numerical example

- Suppose standard functional forms

$$f(k) = k^\alpha, \quad u(c) = \frac{c^{1-\sigma} - 1}{1-\sigma}$$

with parameter values  $\alpha = 0.3$ ,  $\sigma = 1$  and  $\rho = \delta = 0.05$  (annual)

- Implies steady state values

$$\frac{\bar{k}}{\bar{y}} = \frac{\alpha}{\rho + \delta} = 3, \quad \frac{\bar{c}}{\bar{y}} = \frac{\rho + (1 - \alpha)\delta}{\rho + \delta} = 0.85$$

and

$$f''(\bar{k})\bar{k} = -(1 - \alpha)(\rho + \delta) = -0.07$$

## Numerical example

- Use these steady state values to get coefficient matrix

$$\mathbf{A} = \begin{pmatrix} 1 - \frac{\beta f''(\bar{k})\bar{c}}{\sigma} & \frac{f''(\bar{k})\bar{k}}{\sigma} \\ -\frac{\bar{c}}{\bar{k}} & \frac{1}{\beta} \end{pmatrix} = \begin{pmatrix} 1.02 & -0.07 \\ -0.28 & 1.05 \end{pmatrix}$$

- Trace

$$\text{tr}(\mathbf{A}) = 1.02 + 1.05 = 2.07$$

- Determinant

$$\det(\mathbf{A}) = (1.02)(1.05) - (0.07)(0.28) = 1.05$$

## Numerical example

- Eigenvalues are roots of characteristic polynomial

$$p(\lambda) = \lambda^2 - \text{tr}(\mathbf{A})\lambda + \det(\mathbf{A}) = \lambda^2 - (2.07)\lambda + 1.05 = 0$$

- Solving this gives

$$\lambda_1 = 0.89, \quad \lambda_2 = 1.18$$

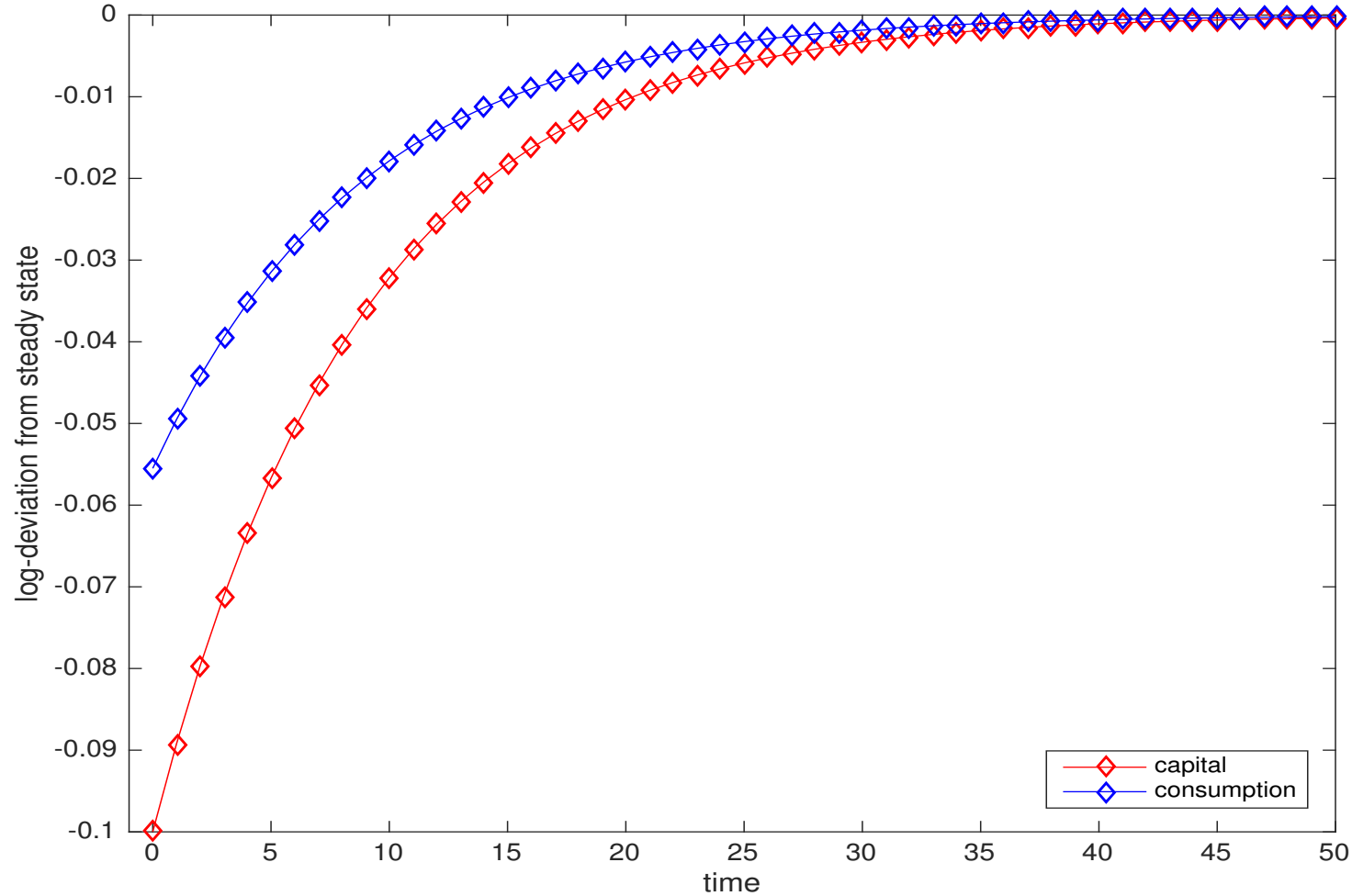
- Choose stable root

$$\psi_{kk} = \lambda_1 = 0.89$$

- Implies slope of the stable arm

$$\psi_{ck} = \left( \frac{1}{\beta} - \psi_{kk} \right) \frac{\bar{k}}{\bar{c}} = \frac{1.05 - 0.89}{0.28} = 0.56$$

# Transition to steady state



Initial capital  $\hat{k}_0 = -0.1$  (i.e., 10% below steady state). Capital  $\hat{k}_{t+1} = \psi_{kk} \hat{k}_t$  and consumption  $\hat{c}_t = \psi_{ck} \hat{k}_t$  with  $\psi_{kk} = 0.89$  and  $\psi_{ck} = 0.56$ .

# Scraps of Matlab code

From Matlab script “*optimal\_growth\_example.m*” in LMS

```
%%%% parameters

alpha = 0.30;      %% capital's share
beta  = 1/1.05;    %% discount factor
delta = 0.05;      %% depreciation rate
sigma = 1;         %% coefficient relative risk aversion
```

```

%%%%% steady state

rho    = 1/beta - 1;           %% discount rate

r      = rho+delta;           %% steady-state mpk

kbar   = (alpha/r)^(1/(1-alpha)); %% steady-state k
ybar   = kbar^alpha           ; %% steady-state y
cbar   = ybar - delta*kbar;    %% steady-state c

d2f    = alpha*(alpha-1)*kbar^(alpha-2); %% f''(kbar)

```

```
%%%%% coefficient matrix
```

```
AA = [ 1-beta*d2f*cbar/sigma , d2f*kbar/sigma;  
      -cbar/kbar              , 1/beta      ];
```

```
%%%%% eigenvalues are roots of quadratic in lambda
```

```
lambdas = roots([1,-trace(AA),det(AA)]);
```

```

%%%%%%%% choose stable root

lambda = min(abs(lambdas));

if abs(lambda)>1,
    display('check roots')
end

%%%%%%%% solution

psi_kk = lambda;
psi_ck = (1/beta - psi_kk)*kbar/cbar;

```



## Transition path from initial condition

```
%%%%% transitional dynamics

k0      = -0.2; %% initial k, log-dev from steady state

T       = 51;   %% horizon

kt      = zeros(T,1);
ct      = zeros(T,1);

kt(1)   = k0;
ct(1)   = psi_ck*k0;

for t=1:T-1,

    kt(t+1) = psi_kk*kt(t);
    ct(t+1) = psi_ck*kt(t+1);

end
```

## Simple plot

```
time = (0:1:T-1); %% time index

figure(1)
plot(time, kt, 'rd-', time, ct, 'bd-')
ylabel('log-deviation from steady state')
xlabel('time')
axis([min(time) max(time) k0 0])
legend('capital', 'consumption', 'location', 'SouthEast')
```

## Next class

- Ramsey-Cass-Koopmans growth model in continuous time
  - brief introduction to optimal control theory
  - decentralization of planning problem