

# **Macroeconomics A: Review Session IV**

Solow Growth Model

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# Outline

## **1 Understanding the Solow Growth Model**

- Capital Accumulation
- Speed of Convergence

## **2 Adding Human Capital to the Solow Model**

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## **1 Understanding the Solow Growth Model**

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## 2 Adding Human Capital to the Solow Model

# Solow Growth Model

- Solow growth model has a simple premise

$$\dot{k}(t) = s_k y(t) - (\delta + n + g)k(t)$$

- When  $\dot{k} = 0$  (steady state), the share of output going to capital is equal to capital depreciation plus the growth rate of effective labor
- Given that  $y(t) = (k(t))^\alpha$

$$k^* = \left( \frac{s_k}{\delta + n + g} \right)^{\frac{1}{1-\alpha}}$$

- This level of capital maximizes consumption
- As we will see, no need for utility function to pin equilibrium  $s_k$  down

# Finding the Capital Law of Motion

- To find  $\dot{k}(t)$  remember it is scaled capital

$$k(t) = \frac{K(t)}{A(t)L(t)}$$

- Taking the derivative with respect to time gives

$$\frac{dk(t)}{dt} = \frac{1}{A(t)L(t)} \frac{\partial K_t}{\partial t} - \underbrace{\frac{K(t)}{A(t)(L(t))^2} \frac{\partial L(t)}{\partial t}}_{k(t) \frac{1}{L(t)} \frac{\partial L(t)}{\partial t}} - \underbrace{\frac{K(t)}{(A(t))^2 L(t)} \frac{\partial A_t}{\partial t}}_{k(t) \frac{1}{A(t)} \frac{\partial A(t)}{\partial t}}$$

- We assume that labor and productivity have constant growth rates

$$\frac{1}{L(t)} \frac{\partial L(t)}{\partial t} = \frac{\dot{L}(t)}{L(t)} = n \quad \frac{1}{A(t)} \frac{\partial A(t)}{\partial t} = \frac{\dot{A}(t)}{A(t)} = g$$

# Derivation

- Combining the result on the previous slide

$$\dot{k}(t) = \frac{\dot{K}(t)}{A(t)L(t)} - (n + g)k(t)$$

- By definition, the change in capital is equal to total saving less depreciation

$$\dot{K}(t) = s_K Y(t) - \delta K(t)$$

- Using this identity

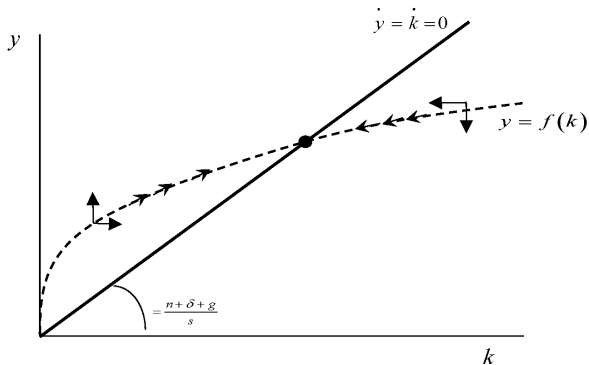
$$\dot{k}(t) = \frac{s_K Y(t) - \delta K(t)}{A(t)L(t)} - (n + g)k(t)$$

$$\implies \dot{k}(t) = s_K y(t) - (\delta + n + g)k(t)$$

# Saddle Path

Figure: Transition Dynamics in the Solow Growth Model

A semi-Phase diagram for the Solow model



# Off Equilibrium

- Almost by definition, a higher saving rate should increase  $k^*$  and  $y^*$

$$\frac{\partial k^*}{\partial s_k} = \frac{1}{(1-\alpha)(\delta+n+g)} \left( \frac{s_k^*}{\delta+n+g} \right)^{\frac{1}{1-\alpha}-1} = \frac{1}{1-\alpha} \frac{k^*}{s_k} > 0$$

- However, this does not mean consumption increases

$$c^* = (k^*)^\alpha - (\delta+n+g)k^*$$

Exercise: find  $\frac{\partial c^*}{\partial s_k}$ . Under what condition is  $c^*$  maximized?

Hint: recall that  $k^* = \left( \frac{s_k}{\delta+n+g} \right)^{\frac{1}{1-\alpha}}$



# Off Equilibrium

Exercise: find  $\frac{\partial c^*}{\partial s_k}$ . Under what condition is  $c^*$  maximized?

$$\begin{aligned}\frac{\partial c^*}{\partial s_k} &= \alpha(k^*)^{\alpha-1} \frac{\partial k^*}{\partial s_k} - (\delta + n + g) \frac{\partial k^*}{\partial s_k} \\ &= \frac{\alpha}{1-\alpha} \frac{(k^*)^\alpha}{s_k} - \frac{1}{1-\alpha} \left( \frac{s_k}{\delta + n + g} \right)^{\frac{\alpha}{1-\alpha}}\end{aligned}$$

Using the definition of  $k^*$

$$\begin{aligned}\frac{\partial c^*}{\partial s_k} &= \frac{\alpha}{1-\alpha} \frac{(k^*)^\alpha}{s_k} - \frac{1}{1-\alpha} (k^*)^\alpha \\ &= \left( \frac{\alpha}{s_k} - 1 \right) \frac{(k^*)^\alpha}{1-\alpha}\end{aligned}$$

Finally, note that

$$s_k = \alpha \quad \implies \quad \frac{\partial c^*}{\partial s_k} = 0$$

# Finding the Speed of Convergence

- We know the law of motion for capital and would like to make it an explicit function of time

$$\dot{k}(t) = s_k k(t)^\alpha - (\delta + n + g)k(t)$$

- Having the path of capital allows us to solve output at  $t$

$$\frac{\dot{y}(t)}{y(t)} = \alpha \frac{\dot{k}(t)}{k(t)}$$

- One option is a 1st order Taylor approximation around  $k(t) = k^*$

$$\begin{aligned}\dot{k}(t) &\approx \left[ \alpha s_k (k^*)^{\alpha-1} - (\delta + n + g) \right] (k(t) - k^*) \\ &\approx (\alpha - 1)(\delta + n + g)(k(t) - k^*)\end{aligned}$$

# Nonhomogenous Differential Equation

- When  $\dot{x} - ax(t) = b$ , we can write

$$x(t) = -\frac{b}{a} + \left( x(0) + \frac{b}{a} \right) e^{at}$$

- The Taylor approximation on the previous slide gives

$$\dot{k}(t) - \underbrace{(\alpha - 1)(\delta + n + g)}_a k(t) = \underbrace{(1 - \alpha)(\delta + n + g)}_b k^*$$

- So we can write

$$k(t) = -\frac{b}{a} + \left( k(0) + \frac{b}{a} \right) e^{at}$$

# The Effect of $\alpha$ on Convergence

- We can now write  $k(t)$  as an explicit function of  $t$

$$k(t) - k^* = (k(0) - k^*) \exp[(\alpha - 1)(\delta + n + g)t]$$

- Taking some interval  $c = \frac{k(0) - k^*}{k(t) - k^*} > 1$  (assuming  $k(t) \rightarrow k^*$ )

$$\frac{1}{c} = \exp[(\alpha - 1)(\delta + n + g)t]$$

$$t = \frac{\log(c)}{(1 - \alpha)(\delta + n + g)}$$

- Note that  $t$  is increasing in  $\alpha$  and decreasing in the other parameters
- Is this intuitive? What does  $\alpha$  represent?

Excel model is [here](#)

# Lack of Convergence?

GDP per person (US = 1) in 2011

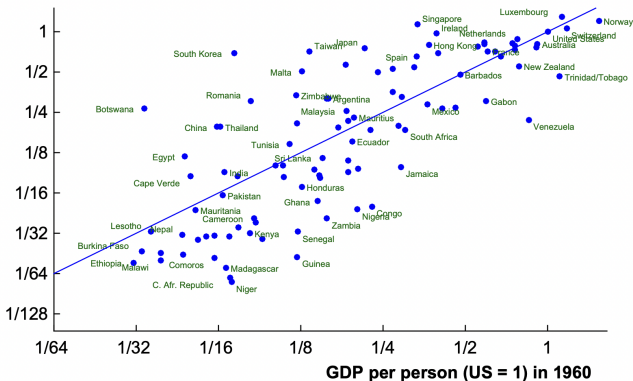


Fig. 24 GDP per person, 1960 and 2011. Source: *The Penn World Tables 8.0*.

From C.I. Jones excellent overview [The Facts of Economic Growth](#)

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# Human Capital and Speed of Convergence

- We can add a human capital stock to the production function

$$y(t) = k(t)^\alpha h(t)^\beta$$

- The law of motion for  $k(t)$  and  $h(t)$  are similar

$$\dot{k}(t) = s_k y(t) - (\delta + n + g)k(t)$$

$$\dot{h}(t) = s_h y(t) - (\delta + n + g)h(t)$$

- The growth rate of  $y$  is given by  $\partial y / \partial t$

$$\frac{\dot{y}(t)}{y(t)} = \alpha \frac{\dot{k}(t)}{k(t)} + \beta \frac{\dot{h}(t)}{h(t)} \quad (1)$$

- To find the speed of convergence, we need to find  $\frac{\dot{k}(t)}{k(t)}$  and  $\frac{\dot{h}(t)}{h(t)}$

# Taylor Approximation Around Equilibrium

- Let's take a Taylor approximation around  $k(t) = k^*$  and  $h(t) = h^*$
- For  $\dot{k}(t)/k(t)$

$$\begin{aligned}f(k(t), h(t)) &= f(k^*, h^*) + f_k(k^*, h^*)(k(t) - k^*) + f_h(k^*, h^*)(h(t) - h^*) \\&= (\alpha - 1)s_k \frac{y^*}{(k^*)^2} (k(t) - k^*) + \beta s_k \frac{y^*}{k^* h^*} (h(t) - h^*) \\&= (\delta + n + g) \left[ (\alpha - 1) \frac{k(t) - k^*}{k^*} + \beta \frac{h(t) - h^*}{h^*} \right]\end{aligned}$$

- For  $\dot{h}(t)/h(t)$

$$\begin{aligned}f(k(t), h(t)) &= f(k^*, h^*) + f_k(k^*, h^*)(k(t) - k^*) + f_h(k^*, h^*)(h(t) - h^*) \\&= \alpha s_h \frac{y^*}{h^* k^*} (k(t) - k^*) + (1 - \beta) s_h \frac{y^*}{(h^*)^2} (h(t) - h^*) \\&= (\delta + n + g) \left[ \alpha \frac{k(t) - k^*}{k^*} + (\beta - 1) \frac{h(t) - h^*}{h^*} \right]\end{aligned}$$



# Taylor Approximation of Output

- The notation is getting heavy, so let's specify a convenience variable

$$\theta(t) = \left[ \alpha \frac{k(t) - k^*}{k^*} + \beta \frac{h(t) - h^*}{h^*} \right]$$

- Going back to equation 1, we can now write this as

$$\frac{\dot{y}(t)}{y^*} \approx (\alpha + \beta - 1)(\delta + n + g)\theta(t)$$

- A Taylor approximation of  $y(t)$  at  $y(t) = y^*$  gives

$$\begin{aligned} y(t) &\approx y^* + \alpha y^* \frac{k(t) - k^*}{k^*} + \beta y^* \frac{h(t) - h^*}{h^*} \\ \implies \frac{y(t) - y^*}{y^*} &= \theta(t) \end{aligned}$$

# Using a Differential Equation

- The law of motion for  $y(t)$  is now given by

$$\dot{y}(t) \approx (\alpha + \beta - 1)(\delta + n + g)(y(t) - y^*)$$

- Using the same solution as before

$$y(t) - y^* = (y(0) - y^*) \exp[-(1 - \alpha - \beta)(\delta + n + g)t]$$

- Taking some interval  $c = \frac{y(0) - y^*}{y(t) - y^*}$

$$\frac{1}{c} = \exp[-(1 - \alpha - \beta)(\delta + n + g)t]$$

$$t = \frac{\log(c)}{(1 - \alpha - \beta)(\delta + n + g)}$$

- Note that  $t$  is increasing in both  $\alpha$  and  $\beta$ , decreasing in the other parameters