

# Lecture 2: Growth Model, Dynamic Optimization in Discrete Time

ECO 503: Macroeconomic Theory I

Benjamin Moll

Princeton University

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## Recall from last lecture

- Economy  $\Leftrightarrow$  resource allocation problem  $\Leftrightarrow$  primitives
- where primitives =
  - preferences
  - technology
  - endowments
- This lecture: economy = growth model
- Next slide: complete description of economy in terms of primitives

## Growth Model: Setup

- **Preferences:** a single household with preferences defined by

$$\sum_{t=0}^{\infty} \beta^t u(c_t, 1 - h_t)$$

with  $u : \mathbb{R}_+ \times [0, 1] \rightarrow \mathbb{R}$

- **Technology:**

$$\begin{aligned}y_t &= F(k_t, h_t), & F : \mathbb{R}_+ \times \mathbb{R}_+ &\rightarrow \mathbb{R}_+ \\c_t + i_t &= y_t \\k_{t+1} &= i_t + (1 - \delta)k_t \\c_t &\geq 0, & i_t &\geq -(1 - \delta)k_t\end{aligned}$$

- **Endowments:**

- 1 unit of time each period
- $\hat{k}_0$  units of capital at time 0

## Assumptions

- **Preferences:**  $0 < \beta < 1$  and  $u$  is
  - strictly increasing
  - strictly concave
  - $C^2$  (twice continuously differentiable)
- **Technology:**  $0 < \delta \leq 1$  and  $F$  is
  - constant returns to scale
  - strictly increasing
  - weakly concave in  $(k, h)$  jointly, strictly concave in each argument individually
  - $F(0, h) = 0$  for all  $h$ .
  - $C^2$
  - (“Inada conditions”)

$$\lim_{k \rightarrow 0} F_k(k, h) = \infty, \quad \forall h > 0,$$

$$\lim_{k \rightarrow \infty} F_k(k, h) = 0, \quad \forall h > 0,$$

## Comments

- **Tradeoffs** in the model
  - consumption today  $c_t$  vs. consumption tomorrow  $c_{t+1}$
  - consumption  $c_t$  vs. leisure  $1 - h_t$
- Model assumes **“representative household”** and **“representative firm”** (jointly = “representative agent”)
- When is this justified? If at least one of following 3 conditions are satisfied
  - ① all individuals in economy are identical
  - ② particular assumptions on preferences (“homotheticity”, “Gorman aggregation”)
  - ③ perfect markets
    - representative firm  $\Leftrightarrow$  perfect factor markets (capital, labor), equalize marginal products
    - representative HH  $\Leftrightarrow$  perfect insurance markets, equalize marginal utilities
- Do we believe these conditions are satisfied? **No**, but...

## General Comment: Modeling in (Macro)economics

- Objective is **not** to build one big model that we use to address all issues
  - descriptive realism is not the objective
  - instead make modeling choices that are dependent on the issue
  - whether a model is “good” is context dependent
- Approach to modeling in macro(economics) well summarized by following two statements
  - “All models are false; some are useful”
  - “If you want a model of the real world, look out the window” (kidding, but only half kidding)

## General Comment: Modeling in (Macro)economics

- But: growth model is **“the”** benchmark model of macro
- Why is this the benchmark model?
  - minimal model of  $y$  where  $y = F(k, h)$
- Also, growth model = great laboratory for teaching you tools of macro...
- ... and many other models in macroeconomics build on growth model. Examples:
  - Real business cycle (RBC) model = growth model with aggregate productivity shocks
  - New Keynesian model = RBC model + sticky prices
  - Incomplete markets model (Aiyagari-Bewley-Huggett) = growth model + heterogeneity in form of uninsurable idiosyncratic shocks

## What issues is growth model useful for?

- Growth model is designed to be model of capital accumulation process
- Growth model is **not** a “good” model of
  - growth (somewhat ironically given its name)
  - income and wealth distribution (given rep. agent assumption)
  - inflation and monetary policy
  - unemployment
  - financial crises
- But some of growth model’s extensions (e.g. those mentioned on previous slide) are “good” models of these issues



## Some Concepts

- **Definition:** A **feasible allocation** for the growth model is a list of sequences  $\{c_t, h_t, k_t\}$  such that

$$c_t + k_{t+1} = F(k_t, h_t) + (1 - \delta)k_t$$

$$0 \leq h_t \leq 1, \quad c_t \geq 0, \quad k_t \geq 0, \quad k_0 = \hat{k}_0$$

# Analysis of Growth Model

- Consistent with there being two key tradeoffs, captured by the model, there are two choices to be made each period
  - $c_t$  vs.  $k_{t+1}$
  - $c_t$  vs.  $h_t$
- Will analyze
  - ① Pareto efficient allocations
  - ② decentralized equilibrium allocations
- Start with Pareto efficient allocations

## Solow Model

- Historically, much interest in allocations that resulted from specific “ad hoc” decision rules

$$c_t = sy_t$$

$$h_t = \bar{h}$$

- = “Solow model” you may know from your undergraduate courses
- See homework 1

## Pareto Efficient Alloc. in Growth Model

- To simplify analysis and focus on dynamics considerations, begin with extreme case: leisure not valued, or (with slight abuse of notation)

$$u(c_t, 1 - h_t) = u(c_t)$$

- Assume (Inada condition akin to those on  $F$ )

$$\lim_{c \rightarrow 0} u'(c) = \infty$$

- Also define

$$f(k_t) = F(k_t, 1)$$

## Social Planner's Problem

- Only one person in economy  $\Rightarrow$  our life is simple.
- Pareto efficient allocation = max. utility of household subject to feasibility
- Think of this as problem of fictitious “social planner”:

$$V(\hat{k}_0) = \max_{\{c_t, k_{t+1}\}_{t=0}^{\infty}} \sum_{t=0}^{\infty} \beta^t u(c_t) \quad \text{s.t.}$$
$$c_t + k_{t+1} = f(k_t) + (1 - \delta)k_t$$
$$c_t \geq 0, \quad k_{t+1} \geq 0, \quad k_0 = \hat{k}_0.$$

- Alternatively, substitute resource constraint into objective

$$V(\hat{k}_0) = \max_{\{k_{t+1}\}_{t=0}^{\infty}} \sum_{t=0}^{\infty} \beta^t u(f(k_t) + (1 - \delta)k_t - k_{t+1}) \quad \text{s.t.}$$
$$0 \leq k_{t+1} \leq f(k_t) + (1 - \delta)k_t, \quad k_0 = \hat{k}_0.$$

# Dynamic Optimization: General Theory

- There's a general theory for solving these types of problems
  - let's first work out more general theory
  - then apply to growth model
  - purpose: teach you some tools that are also applicable for solving other models
- In general will encounter two different formulations of dynamic optimization problems
  - ① control-state formulation
  - ② state-only formulation
- see previous slide, return to this momentarily

# Dynamic Optimization: General Theory

## Control-State Formulation

- Recall discussion of two formulations
  - do state-control formulation first
  - then do state-only formulation
- Pretty much all deterministic optimal control problems in discrete time can be written as

$$V(\hat{x}_0) = \max_{\{z_t\}_{t=0}^{\infty}} \sum_{t=0}^{\infty} \beta^t h(x_t, z_t)$$

subject to the law of motion for the state

$$x_{t+1} = g(x_t, z_t) \text{ and } z_t \in Z, \quad x_0 = \hat{x}_0.$$

- $\beta \in (0, 1)$ : discount factor
- $x \in X \subseteq \mathbb{R}^m$ : state vector
- $z \in Z \subseteq \mathbb{R}^k$ : control vector
- $h : X \times Z \rightarrow \mathbb{R}$ : instantaneous return function

## Example: Growth Model

$$V(\hat{k}_0) = \max_{\{c_t, k_{t+1}\}_{t=0}^{\infty}} \sum_{t=0}^{\infty} \beta^t u(c_t) \quad \text{s.t.}$$
$$c_t + k_{t+1} = f(k_t) + (1 - \delta)k_t$$
$$c_t \geq 0, \quad k_{t+1} \geq 0, \quad k_0 = \hat{k}_0.$$

- Here the state is  $x_t = k_t$  and the control  $z_t = c_t$
- $h(x, z) = u(z)$
- $g(x, z) = f(x) + (1 - \delta)x - z$



# Dynamic Optimization: General Theory

## State-only Formulation

- Alternatively, can write the same problem in terms of states only

$$V(\hat{x}_0) = \max_{\{x_{t+1}\}_{t=0}^{\infty}} \sum_{t=0}^{\infty} \beta^t U(x_t, x_{t+1}) \quad \text{s.t.}$$
$$x_{t+1} \in \Gamma(x_t), \quad x_0 = \hat{x}_0.$$

- $\beta \in (0, 1)$ : discount factor
- $x \in X \subseteq \mathbb{R}^m$ : state vector
- $U : X \times X \rightarrow \mathbb{R}$ : instantaneous return function
- $\Gamma : X \rightarrow X$ : correspondence describing feasible values for state

## Example: Growth Model

$$V(\hat{k}_0) = \max_{\{k_{t+1}\}_{t=0}^{\infty}} \sum_{t=0}^{\infty} \beta^t u(f(k_t) + (1 - \delta)k_t - k_{t+1}) \quad \text{s.t.}$$
$$k_{t+1} \in [0, f(k_t) + (1 - \delta)k_t], \quad k_0 = \hat{k}_0.$$

- Here the state is  $x_t = k_t$
- $U(x, y) = f(x) + (1 - \delta)x - y$
- $\Gamma(x) = [0, f(x) + (1 - \delta)x]$

# Dynamic Optimization: General Properties

- **Existence** of a solution
  - Extreme Value Theorem (or “Weierstrass Theorem”): continuous function on compact set has a maximum
- Satisfied in growth model?
  - objective continuous? Yes
  - constraint set compact? Yes. Result: there exists a “maximum maintainable capital stock”  $\hat{k}$  s.t.  $k_t > \hat{k} \Rightarrow k_{t+1} - k_t < 0$ , and we can restrict attention to  $k_t \in [0, \hat{k}]$ .
    - Inada conditions  $\Rightarrow f'(k_t) - \delta < 0$  for  $k_t$  large enough  $\Rightarrow$  there exists  $\hat{k}$  satisfying  $0 = f(\hat{k}) - \delta\hat{k}$  and  $f(k_t) - \delta k_t < 0, k_t > \hat{k}$
    - $k_t > \hat{k} \Rightarrow k_{t+1} - k_t = f(k_t) - \delta k_t - c_t \leq f(k_t) - \delta k_t < 0$
  - $\Rightarrow$  in growth model, there exists an optimal  $\{k_{t+1}\}_{t=0}^{\infty}$
- **Uniqueness** of a solution
  - strictly concave objective & convex constraint set  
 $\Rightarrow$  unique solution
- Satisfied in growth model? Yes

## Overview: Solution Methods

- There are different methods for solving dynamic optimization problems
  - not only deterministic ones ...
  - ... but also stochastic ones (= with uncertainty)
  - Table provides an overview of different solution methods

	Discrete	Time	Continuous	Time
	sequence	recursive	sequence	recursive
deterministic	“classical”	Bellman eqn	Hamiltonian	HJB eqn
stochastic		Bellman eqn		HJB eqn

- blue = this class (first six weeks of 503)
- recursive approach also called “dynamic programming”
- blank box
  - **can** solve stochastic problems using sequence formulation...
  - ... but recursive/dynamic programming approach strictly

## Classical Solution Method of Sequence Pb.

- Recall general dynamic optimization problem

$$V(\hat{x}_0) = \max_{\{x_{t+1}\}_{t=0}^{\infty}} \sum_{t=0}^{\infty} \beta^t U(x_t, x_{t+1}) \quad \text{s.t.} \quad (P)$$
$$x_{t+1} \in \Gamma(x_t), \quad x_0 = \hat{x}_0.$$

- The following are necessary and sufficient conditions for  $\{x_{t+1}\}_{t=0}^{\infty}$  to be optimal: if  $x_{t+1}$  is in the interior of  $\Gamma(x_t)$

$$U_y(x_t, x_{t+1}) + \beta U_x(x_{t+1}, x_{t+2}) = 0, \quad \forall t \quad (EE)$$

$$\lim_{T \rightarrow \infty} \beta^T U_y(x_T, x_{T+1}) \cdot x_{T+1} = 0 \quad (TC)$$

and  $x_0 = \hat{x}_0$ .

- (EE) together with (TC) and initial condition  $x_0 = \hat{x}_0$  fully characterizes optimal  $\{x_{t+1}\}_{t=0}^{\infty}$

## Derivation/Interpretation

- (EE) is called “Euler equation”
  - simply first-order condition (FOC) w/ respect to  $x_{t+1}$
  - derivation: differentiate problem (P) with respect to  $x_{t+1}$
  - “Euler equation” simply means “intertemporal FOC”
- (TC) is called “transversality condition”
  - understanding it is harder than (EE), let’s postpone this for now and revisit in a few slides
  - Note: some books (e.g. Stokey-Lucas-Prescott) write (TC) as

$$\lim_{T \rightarrow \infty} \beta^T U_x(x_T, x_{T+1}) \cdot x_T = 0 \quad (\text{TC2})$$

- To see that (TC2) is equivalent to (TC), substitute (EE) into (TC), and evaluate at  $T$  rather than  $T + 1$

## Example: Growth Model

- Recall social planner's problem in growth model

$$V(\hat{k}_0) = \max_{\{k_{t+1}\}_{t=0}^{\infty}} \sum_{t=0}^{\infty} \beta^t u(f(k_t) + (1 - \delta)k_t - k_{t+1}) \quad \text{s.t.} \quad (\text{P}')$$

$$k_{t+1} \in [0, f(k_t) + (1 - \delta)k_t], \quad k_0 = \hat{k}_0.$$

- (EE) and (TC) are

$$\begin{aligned} & -u'(f(k_t) + (1 - \delta)k_t - k_{t+1}) && (\text{EE}') \\ & + \beta u'(f(k_{t+1}) + (1 - \delta)k_{t+1} - k_{t+2})(f'(k_{t+1}) + 1 - \delta) = 0 \end{aligned}$$

$$\lim_{T \rightarrow \infty} \beta^T u'(f(k_T) + (1 - \delta)k_T - k_{T+1})k_{T+1} = 0 \quad (\text{TC}')$$

- Get (EE') simply by differentiating (P') w.r.t.  $k_{t+1}$  (or by applying formula on previous slide)

## Example: Growth Model

- (EE') can be written more intuitively as

$$\underbrace{\frac{u'(c_t)}{\beta u'(c_{t+1})}}_{MRS} = \underbrace{f'(k_{t+1}) + 1 - \delta}_{MRT}$$

MRS between  $c_t$  and  $c_{t+1}$  = MRT between  $c_t$  and  $c_{t+1}$

- Same logic as in static utility maximization problems, e.g.

$$\max_{c_A, c_B} u(c_A, c_B) \quad \text{s.t.} \quad c_A = f(\ell_A), \quad c_B = f(\ell_B), \quad \ell_A + \ell_B \leq 1$$

where  $A$ =apples,  $B$ =bananas

$$\Rightarrow \frac{u_{c_A}(c_A, c_B)}{u_{c_B}(c_A, c_B)} = \frac{f'(\ell_A)}{f'(\ell_B)}$$

- growth model: different dates = different goods



## Example: Growth Model

- Summarizing all necessary conditions

$$\begin{aligned}u'(c_t) &= \beta u'(c_{t+1})(f'(k_{t+1}) + 1 - \delta) \\k_{t+1} &= f(k_t) + (1 - \delta)k_t - c_t\end{aligned}\tag{DE}$$

for all  $t$ , with

$$\begin{aligned}k_0 &= \hat{k}_0 \\ \lim_{T \rightarrow \infty} \beta^T u'(c_T) k_{T+1} &= 0\end{aligned}\tag{TC'}$$

- (DE) is system of two difference equations in  $(c_t, k_t)$  ...
- ... needs two boundary conditions
  - initial condition for capital stock:  $k_0 = \hat{k}_0$
  - transversality condition, **plays role of boundary condition**

## Where does (TC) come from?

- Transversality condition is a bit mysterious
- Best treatments are in various papers by Kamihigashi
  - most intuitive “Transversality Conditions and Dynamic Economic Behavior,” New Palgrave Dict. of Economics, 2008  
[http://www.dictionaryofeconomics.com/download/pde2008\\_T000217.pdf](http://www.dictionaryofeconomics.com/download/pde2008_T000217.pdf)
  - “A simple proof of the necessity of the transversality condition,” Economic Theory, 2002
  - “Necessity of transversality conditions for infinite horizon problems,” Econometrica, 2001
- Next slide: intuitive but “fake” derivation from finite horizon problem
- Afterwards: necessity proof from Kamihigashi (2002)

## Where does (TC) come from?

- Consider finite horizon problem:

$$V(\hat{k}_0, T) = \max_{\{k_{t+1}\}_{t=0}^T} \sum_{t=0}^T \beta^t u(c_t) \quad \text{s.t.}$$
$$k_{t+1} = f(k_t) + (1 - \delta)k_t - c_t, \quad k_{t+1} \geq 0.$$

- Lagrangian

$$\mathcal{L} = \sum_{t=0}^T \beta^t u(c_t) + \sum_{t=0}^T \lambda_t (f(k_t) + (1 - \delta)k_t - c_t - k_{t+1}) + \sum_{t=0}^T \mu_t k_{t+1}$$

- Necessary conditions at  $t = T$

$$\begin{aligned} \beta^T u'(c_T) &= \lambda_T \\ \lambda_T &= \mu_T \quad \Rightarrow \quad \beta^T u'(c_T) k_{T+1} = 0 \\ \mu_T k_{T+1} &= 0 \end{aligned}$$

## Where does (TC) come from?

- From previous slide: in finite horizon problem

$$\beta^T u'(c_T) k_{T+1} = 0 \quad (*)$$

- (\*) is really two conditions in one

- ①  $\beta^T u'(c_T) > 0$ : need  $k_{T+1} = 0$
- ②  $\beta^T u'(c_T) = 0$ :  $k_{T+1}$  can be  $> 0$

- Intuition for case 1: if my marginal utility of consumption at  $T$  is positive, I want to eat up all my wealth before I die
- (TC) is same condition as (\*) in economy with  $T \rightarrow \infty$

$$\lim_{T \rightarrow \infty} \beta^T u'(c_T) k_{T+1} = 0 \quad (\text{TC})$$

- Intuition:
  - capital should not grow too fast compared to marginal utility
  - e.g. with  $u(c) = \log c$ :  $\beta^T k_{T+1}/c_{T+1} \rightarrow 0$
  - if I save too much/spend too little, I'm not behaving optimally
- (TC) rules out **overaccumulation** of wealth

# Proof of Necessity of (TC)

Kamihigashi (2002)

- Consider general optimal control problem

$$V(\hat{x}_0) = \max_{\{x_{t+1}\}_{t=0}^{\infty}} \sum_{t=0}^{\infty} \beta^t U(x_t, x_{t+1}) \quad \text{s.t.} \quad (P)$$
$$x_{t+1} \in \Gamma(x_t), \quad x_0 = \hat{x}_0.$$

- **Assumptions:**

- ①  $x_t \in X \subset \mathbb{R}_+^m$  (i.e.  $x_t \geq 0$ )
  - ②  $Gr(\Gamma) = \{(y, x) : x \in X, y \in \Gamma(x)\}$  is convex,  $(0, 0) \in Gr(\Gamma)$
  - ③  $U : Gr(\Gamma) \rightarrow \mathbb{R}$  is  $C^1$  and concave
  - ④  $\forall (x, y) \in Gr(\Gamma), U_y(x, y) \leq 0$
  - ⑤ For any feasible path  $\{x_t\}$   $\lim_{T \rightarrow \infty} \sum_{t=0}^T \beta^t U(x_t, x_{t+1})$  exists (i.e. it is bounded)
- (TC) can also be derived under weaker assumptions. But above assumptions yield straightforward proof.

# Proof of Necessity of (TC)

Kamihigashi (2002)

- **Definition:** A feasible path  $\{x_t^*\}$  is **optimal** if

$$\sum_{t=0}^{\infty} \beta^t U(x_t^*, x_{t+1}^*) \geq \sum_{t=0}^{\infty} \beta^t U(x_t, x_{t+1})$$

for any feasible path  $\{x_t\}$

- i.e.  $\{x_t^*\}$  attains the maximum of (P)
- **Theorem:** Under Assumptions 1 to 5, for any **interior** optimal path  $\{x_t^*\}$

$$\lim_{T \rightarrow \infty} \beta^T U_y(x_T^*, x_{T+1}^*) \cdot x_{T+1}^* = 0$$

## Proof of Necessity of (TC)

Kamihigashi (2002)

- **Useful preliminary fact:** Let  $f : [0, 1] \rightarrow \mathbb{R}$  be a **concave function** with  $f(1) > -\infty$ . Then

$$\frac{f(1) - f(\lambda)}{1 - \lambda} \leq f(1) - f(0) \quad (*)$$

- Kamihigashi calls this a Lemma, not sure it deserves the name
- Follows immediately from definition of a concave function:

$$f(\lambda x + (1 - \lambda)y) \geq \lambda f(x) + (1 - \lambda)f(y) \quad \forall 0 \leq \lambda \leq 1, x, y$$

- Letting  $x = 1$  and  $y = 0$

$$f(1) - f(\lambda) \leq f(1) - (\lambda f(1) + (1 - \lambda)f(0))$$

- Rearranging yields (\*)

# Proof of Necessity of (TC)

Kamihigashi (2002)

- Let  $x_t^*$  be an interior optimal path. Consider alternative path

$$\{x_0^*, x_1^*, \dots, x_T^*, \lambda x_{T+1}^*, \lambda x_{T+2}^*, \dots\}, \quad \lambda \in [0, 1)$$

- path is feasible by interiority and convexity of constraint set
- By optimality

$$\beta^T [U(x_T^*, \lambda x_{T+1}^*) - U(x_T^*, x_{T+1}^*)] + \sum_{t=T+1}^{\infty} \beta^t [U(\lambda x_t^*, \lambda x_{t+1}^*) - U(x_t^*, x_{t+1}^*)] \leq 0$$

- Dividing through by  $1 - \lambda$

$$\begin{aligned} \beta^T \frac{U(x_T^*, \lambda x_{T+1}^*) - U(x_T^*, x_{T+1}^*)}{1 - \lambda} &\leq \sum_{t=T+1}^{\infty} \beta^t \frac{U(x_t^*, x_{t+1}^*) - U(\lambda x_t^*, \lambda x_{t+1}^*)}{1 - \lambda} \\ &\leq \sum_{t=T+1}^{\infty} \beta^t [U(x_t^*, x_{t+1}^*) - U(0, 0)] \end{aligned}$$

where the last inequality follows from A3 (concavity of  $U$ ) and (\*)



# Proof of Necessity of (TC)

Kamihigashi (2002)

- Applying  $\lim_{\lambda \rightarrow 1}$  to the LHS

$$0 \leq -\beta^T U_y(x_T^*, x_{T+1}^*) \cdot x_{T+1}^* \leq \sum_{t=T+1}^{\infty} \beta^t [U(x_t^*, x_{t+1}^*) - U(0, 0)]$$

where the first inequality follows from A4 ( $U_y(x, y) \leq 0$ ) and A1 ( $x_t \geq 0$ )

- Applying  $\lim_{T \rightarrow \infty}$  to both sides

$$\begin{aligned} 0 &\leq -\lim_{T \rightarrow \infty} \beta^T U_y(x_T^*, x_{T+1}^*) \cdot x_{T+1}^* \\ &\leq \lim_{T \rightarrow \infty} \sum_{t=T+1}^{\infty} \beta^t [U(x_t^*, x_{t+1}^*) - U(0, 0)] = 0 \end{aligned}$$

where the equality follows from A5 (boundedness)

- (TC) now follows.  $\square$

## (TC) in Practice

- In practice, often don't have to impose (TC) exactly
- Instead, just have to make sure trajectories "don't blow up."
- E.g. consider growth model: since  $\beta < 1$ , easy to see that

$$\lim_{T \rightarrow \infty} \beta^T u'(c_T) k_{T+1} = 0$$

whenever

$$\lim_{T \rightarrow \infty} c_T = c^*, \quad \lim_{T \rightarrow \infty} k_T = k^*$$

with  $0 < c^*, k^* < \infty$  which is satisfied if  $\{c_t, k_t\}$  converge to steady state.

## Steady State

- **Definition:** a steady state is a point in the state space  $x^*$  such that  $x_0 = x^*$  implies  $x_t = x^*$  for all  $t \geq 1$ . (“if you start there, you stay there”)
- Steady state in general model: any  $x^* \in X$  such that

$$U_y(x^*, x^*) + \beta U_x(x^*, x^*) = 0$$

- Steady state in growth model:  $(c^*, k^*)$  satisfying

$$1 = \beta(f'(k^*) + 1 - \delta) \tag{*}$$

$$c^* = f(k^*) - \delta k^*$$

comes from (DE) with  $c_{t+1} = c_t = c^*$  and  $k_{t+1} = k_t = k^*$

- For example, if  $f(k) = Ak^\alpha$ ,  $\alpha < 1$ . Then

$$k^* = \left( \frac{\alpha A}{\beta^{-1} - 1 + \delta} \right)^{\frac{1}{1-\alpha}}$$

# Dynamics

- What else can we say about dynamics of  $\{c_t\}_{t=0}^{\infty}$  and  $\{k_{t+1}\}_{t=0}^{\infty}$ ?
- Turns out answering this is easier in continuous time
  - phase diagram
  - can also do discrete-time phase diagram, but a bit awkward
  - so rather do it properly
- See next lecture