

# CONTINUOUS STATE VARIABLES: NEOCLASSICAL GROWTH MODEL

---

QUANTITATIVE ECONOMICS 2024

Piotr Żoch

January 10, 2025

## MOTIVATION

- We analyzed finite Markov decision processes (MDP).
  1.  $X$  is a finite set: state space.
  2.  $A$  is a finite set: action space.
- In many interesting applications, state space and action space are not finite.
- Example: a household chooses how much to consume and how much to save: real numbers.
- Instead of studying general results I will focus on a specific example – neoclassical growth model – and illustrate some computational issues.

## NEOCLASSICAL GROWTH MODEL

- The planner maximizes

$$\mathbb{E}_0 \sum_{t=0}^{\infty} \beta^t u(c_t), \quad \text{subject to} \quad k_{t+1} = z_t f(k_t) + (1 - \delta) k_t - c_t \text{ for all } t \geq 0$$

- We restrict  $k_t, c_t \geq 0$  for all  $t \geq 0$ .  $k_0$  is given.
- $\beta \in (0, 1)$  is the discount factor,  $\delta \in (0, 1]$  is the depreciation rate.
- $u$  is a **utility function** and  $f$  is a **production function**.
- $z_t > 0$  is a **productivity shock** that follows a Markov chain with values in  $Z$ . For simplicity assume  $Z$  is finite and  $P$  is a transition matrix.

## NEOCLASSICAL GROWTH MODEL

- We assume
  - $u : \mathbb{R}_+ \rightarrow \mathbb{R}$  is continuous, strictly increasing, strictly concave and twice continuously differentiable with

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

- $f : \mathbb{R}_+ \rightarrow \mathbb{R}$  is continuous, strictly increasing, strictly concave and twice continuously differentiable with

$$\lim_{k \rightarrow 0} f'(k) = \infty, \quad \lim_{k \rightarrow \infty} f'(k) = 0, \quad f(0) = 0.$$

- These assumptions will allow us to ignore corner solutions.

## NEOCLASSICAL GROWTH MODEL

- The state variables are

$$x_t := (k_t, z_t) \in X := \mathbb{R}_+ \times Z.$$

- The state space is **not** finite.
- Let action  $a$  be the choice of capital next period:

$$a_t := k_{t+1} \in A := \mathbb{R}_+.$$

- The action space is **not** finite.
- It is also possible to associate  $a_t$  with consumption  $c_t$ .

## NEOCLASSICAL GROWTH MODEL

- Bellman equation is

$$V(k, z) = \max_{k' \in \Gamma(k, z)} \left\{ u(zf(k) - k') + \beta \sum_{z' \in Z} P(z, z') V(k', z') \right\},$$

where

$$\Gamma(k, z) = [0, zf(k) + (1 - \delta)k].$$

## NEOCLASSICAL GROWTH MODEL

- **Problem:** things we learned about finite state MDPs do not apply here!
- Easier part: theorems that do not require finiteness of state space and action space (see Stokey and Lucas, 1989).
- Harder part: how to solve it numerically in an efficient way?

## DISCRETIZATION



## DISCRETIZATION

- One possible idea is to discretize the state space and action space and then solve the problem numerically using the methods we learned for finite state MDPs.
- We know we will find an exact solution to the **discretized problem**, but will it be a good approximation to the original problem?
- We first need to choose a **grid** for  $k$  and  $z$ .
- We focus on  $k$  – assume that  $z$  is finite state Markov chain (if it is not, we can discretize it as well using Tauchen or Rouwenhurst method).

## DISCRETIZATION

- First step: choose  $k_{\min}$  and  $k_{\max}$  and the number of points  $N_k$ .
- This often requires some experimentation.
  - Too many points – it will make the problem costly to solve.
  - Too few points – it will make the solution to the discretized problem far from the true solution of the original problem.
  - The interval  $[k_{\min}, k_{\max}]$  too wide – it will make the problem costly to solve.
  - The interval  $[k_{\min}, k_{\max}]$  too narrow – it will impose a constraint that was not present in the original problem.

## DISCRETIZATION

- Do not choose  $k_{\min} = 0$ : recall that  $f(0) = 0$ , so if  $\lim_{c \rightarrow 0} u(c) = -\infty$  then you will run into problems.
- To choose  $k_{\max}$ , note that because  $\lim_{k \rightarrow \infty} f'(k) = 0$  for each  $z$  there exists a  $k_{\max,z}$  such that  $zf(k) + (1 - \delta)k < k$  for all  $k > k_{\max,z}$ .

## RESIDUALS

- How to verify that the solution to the discretized problem is a good approximation to the solution of the original problem?
- Recall the RHS of the Bellman equation:

$$\max_{k' \in \Gamma(k, z)} \left\{ u(zf(k) + (1 - \delta)k - k') + \beta \sum_{z' \in Z} P(z, z') V(k', z') \right\}.$$

- The first order condition with respect to  $k'$  is

$$-u'(zf(k) + (1 - \delta)k - k') + \beta \sum_{z' \in Z} P(z, z') V'(k', z') = 0.$$

where  $V'(k', z')$  is the derivative of  $V$  with respect to  $k'$ .

## RESIDUALS

- Evaluate the LHS of the first order condition at the solution to the discretized problem at various  $k$  (including these that did not belong to the grid) and  $z$ .
- If the solution is a good approximation, the LHS should be close to zero.
- What does "close" mean?

## RESIDUALS

- The expression has derivatives of the value function
- We want to have a measure that is easier to interpret.
- Easy to show (envelope condition) that

$$V'(k, z) = (zf'(k) + 1 - \delta) u'(c)$$

where  $c$  is the optimal consumption.

- Calculate

$$\mathcal{H}(k, z) := c - (u')^{-1} \left( \beta \sum_{z' \in Z} P(z, z') z' f'(k') u'(c') \right).$$

## RESIDUALS

- Euler equation residuals are

$$\mathcal{R}(k, z) := \frac{\mathcal{H}(k, z)}{c(k, z)}.$$

- Interpretation: agents make  $\mathcal{R}(k, z)\%$  mistakes when choosing  $c$ .
- Example: if  $\mathcal{R}(k, z)$  is 0.01, then agents spend 1 dollar per 100 dollars "incorrectly".
- Bounded rationality interpretation.

## RESIDUALS

- Euler equation residuals should be quite small.
- They should be even smaller for points  $(k, z)$  that are visited frequently.
- You can immediately see the problem with a grid that is too coarse: grid points are unlikely to correspond to the exact maximizer  $k'$ .



MAXIMIZATION

## BELLMAN EQUATION

- Recall the Bellman equation

$$V(k, z) = \max_{k' \in \Gamma(k, z)} \left\{ u(zf(k) + (1 - \delta)k - k') + \beta \sum_{z' \in Z} P(z, z') V(k', z') \right\},$$

- Solve the maximization problem directly using techniques we learned when we talked about optimization.
- We do not need to assume that  $k'$  belongs to the grid.
- This gives us hope to finding the exact maximizer  $k'$ .
- Problem:** we do not know the exact value of  $V$  at points that do not belong to the grid.

## APPROXIMATION

- We will use an **approximation** of  $V$  at points that do not belong to the grid,  $\hat{V}$ .
- Continuous piecewise linear interpolation is a safe choice.
- Then for each  $k$  and  $z$  we have to solve for  $k'$  using

$$\max_{k' \in \Gamma(k, z)} u(zf(k) + (1 - \delta)k - k') + \beta \sum_{z' \in Z} P(z, z') \hat{V}(k', z').$$

## MODIFICATION

- In the above we have

$$\sum_{z' \in Z} P(z, z') \hat{V}(k', z')$$

- We can approximate  $V$  and then calculate the sum (the expected value)
- It is usually cheaper to approximate the expected value directly.
- Let

$$W(k', z) := \sum_{z' \in Z} P(z, z') V(k', z')$$

- Obtain  $\hat{W}(k', z)$  and then solve

$$\max_{k' \in \Gamma(k, z)} u(zf(k) + (1 - \delta)k - k') + \beta \hat{W}(k', z).$$

## SUMMARY

- Start with some grid, as in the previous section.
- Instead of finding  $k'$  that belongs to the grid (by comparing continuation values directly), solve the maximization problem directly.
- This requires using some optimizer for each  $k$  and  $z$ .
- This requires knowing  $\hat{V}$  at points that do not belong to the grid.
- Important:  $\hat{V}$  must preserve the concavity of  $V$ .

FIRST ORDER CONDITION

## FIRST ORDER CONDITION

- The maximization problem

$$\max_{k' \in \Gamma(k, z)} u(zf(k) + (1 - \delta)k - k') + \beta \sum_{z' \in Z} P(z, z') V(k', z'),$$

has a first order condition

$$-u'(zf(k) + (1 - \delta)k - k') + \beta \sum_{z' \in Z} P(z, z') V'(k', z') = 0.$$

- Instead of solving the maximization problem directly, we can solve the first order condition (a nonlinear equation) for  $k'$ .
- To find it, we can use for example the **bisection method** or **Newton's method**.

## FIRST ORDER CONDITION

- Same comments as with the maximization problem apply,
- We need to approximate  $V'$  at points that do not belong to the grid.
- We can do it by approximating  $V$  and then calculating the derivative or by approximating the derivative directly.
- We can also approximate  $W$  or  $W'$  instead.



## ENDOGENOUS GRID METHOD

## FIRST ORDER CONDITION

- Recall the idea of the previous approach: fix a grid for  $k$  and  $z$  and then solve the nonlinear equation for  $k'$ .
- The nonlinear equation is the most costly part.
- In some cases we can avoid it by using the **endogenous grid method**.

## ENDOGENOUS GRID METHOD

- The first order condition was

$$u'(zf(k) + (1 - \delta)k - k') = \beta W'(k', z)$$

and we wanted to find  $k'$  as a function of  $k$  and  $z$ .

- The endogenous grid method (by Chris Carroll) inverts the logic.
- Suppose you want to go to some  $k'$  given  $z$ . What  $k$  do you need to start with?

## ENDOGENOUS GRID METHOD

- Start with a grid for  $k'$  and  $z$ .
- Invert the first order condition as

$$zf(k) + (1 - \delta)k - k' = (u')^{-1}(\beta W'(k', z))$$

and solve for  $k$ .

- Although you might need to use a nonlinear solver, you do not need to evaluate  $W'$  repeatedly.
- Now we know  $k$  as a function of  $k'$  and  $z$ .
- After we did it for all  $k'$  and  $z$ , we have an endogenous grid for  $k$ .

## ENDOGENOUS GRID METHOD

- We now know  $k(k', z)$  for each  $k'$  and  $z$  in an **exogenous** grid.
- We now use interpolation to get  $k'(k, z)$  for each  $k$  and  $z$  in the same **exogenous** grid.
- After that we have  $V$  for each  $(k, z)$  in the **exogenous** grid.
- Get  $W$  and repeat the process.