

Econ 8185: Quant PS4

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Abstract

The present document use Finite Element Methods to calculate the policy functions.

1 Model

Consider the simple neoclassical growth model with $\delta = 1$.

$$\begin{aligned} \max_{c_t, k_{t+1}} \quad & \sum \beta^t \log(c_t) \\ \text{s.t.} \quad & c_t + k_{t+1} = Ak_t^\alpha \end{aligned}$$

The basic model has an analytical solution for the consumption policy function given by:

$$c(k) = (1 - \beta\alpha)Ak^\alpha.$$

We will approximate the policy function using finite elements methods as outlined below.

1.1 Construction of Linear Basis

We use the following formula to construct linear basis:

$$\psi_i(x) = \begin{cases} \frac{x-x_{i-1}}{x_i-x_{i-1}} & x \in [x_{i-1}, x_i] \\ \frac{x_{i+1}-x}{x_{i+1}-x_i} & x \in [x_i, x_{i+1}] \\ 0 & \text{otherwise} \end{cases}$$

Note: The first and last basis is defined only on one of the relevant intervals.

We plot the basis for grid $X = [0, 1, 3, 6]$ below:

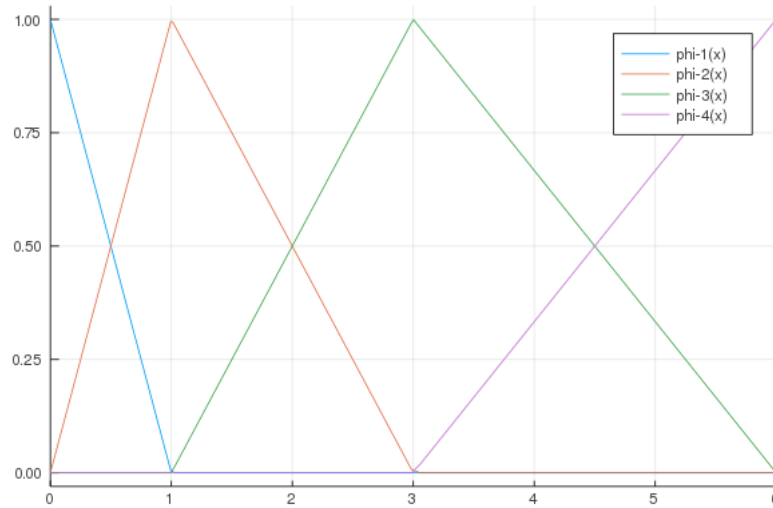


Figure 1: Piecewise Linear Basis Functions

1.2 Estimation of Policy Function using Finite Elements

We approximate the policy function as:

$$c^n(k; \theta) = \sum_i \theta_i \psi_i(k).$$

The residual equation for the present case is:

$$R(k; \theta) = 1 - \beta \frac{c(k'; \theta)}{c(k; \theta)} A \theta (k')^{\alpha-1}$$

$$k' = Ak^\alpha - c(k; \theta)$$

We then use the Glarken Method to evaluate the integral of residual using weights same as the basis function. We set the weighted residual to zero:

$$\int \psi_i(k) R(k; \theta) dk = 0.$$

For $i \in 1, 2, \dots, n$, we have a system of equations. We stack the system and solve for θ using Newton Root finding method for vector valued functions.

Note: We set $\theta_1 = 0$ to ensure $c^n(0; \theta) = 0$. Below is the estimated policy function plotted against the analytical policy function for different grid sizes.

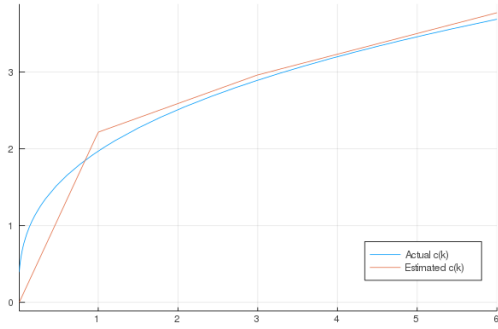


Figure 2: Estimated $c(k)$ with 5 grid points

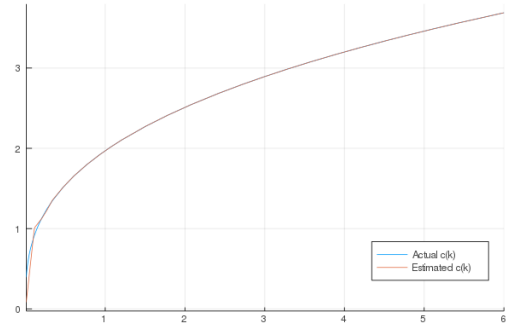


Figure 3: Estimated $c(k)$ with 30 grid points

We note that for decently sized grid, the estimated policy function is almost indistinguishable from the actual policy function for all points except near zero value of capital. To remedy this we can increase the number of grid points around zero. We'll use this idea in estimation of Aiyagari Model.

2 Application to basic Aiyagari Model

The model is:

$$\begin{aligned} \max E_0 \sum \beta^t [c_t^{1-\mu} - 1]/(1-\mu) \\ \text{s.t. } c_t + a_{t+1} = wl_t + (1+r)a_t \\ c_t \geq 0, a_t \geq 0 \end{aligned}$$

where l_t follows a finite state Markov process.

For the present case assume that l_t is iid and takes two values l_L and l_H with equal probabilities. We can take into account the non-negativity constraint by taking into adding a penalty as outlined in the notes. The final Residual equation obtained for this model is:

$$\begin{aligned} R(a, l_t) &= \beta(1+r)E[c'^{-\mu}] + \beta\zeta \min(a', 0)^2 - c^{-\mu} \\ c(a) &= wl_t + (1+r)a - a'(a, l_t) \\ E[c'(a'(a, l_t), l_{t+1})^{-\mu}] &= E_{l_{t+1}}[(wl_{t+1} + (1+r)a'(a, l_t) - a'(a'(a, l_t), l_{t+1}))^{-\mu}] \end{aligned}$$

Since there are two states for l_t , we'll have a asset policy function corresponding to l_L and l_H . We approximate the policy function as follows:

$$\begin{aligned} a^m(a; \theta^L) &= \sum_i \psi_i(a) \theta_i^L \\ a^m(a; \theta^H) &= \sum_i \psi_i(a) \theta_i^H \\ \theta &= [\theta^L, \theta^H]' \end{aligned}$$

As outlined in the previous section we solve a system of equations by setting weighted residuals to zero, where the weights are same as the basis function.

2.1 Results

For the present case we use the following parameter specification (taken from Fran's PS):

$$\begin{aligned} \beta &= 0.9 \\ \mu &= 2 \\ w &= 1.17 \\ r &= 0.04 \\ \zeta &= 30000 \\ l_L &= 0.4, l_H = 1 \end{aligned}$$

We build the asset grid to be between 0 and 20, with 20 exponentially increasing points. We make sure to add more points near zero to account for asset constraint binding. Below we plot the estimated asset policy function for the two states of labor.

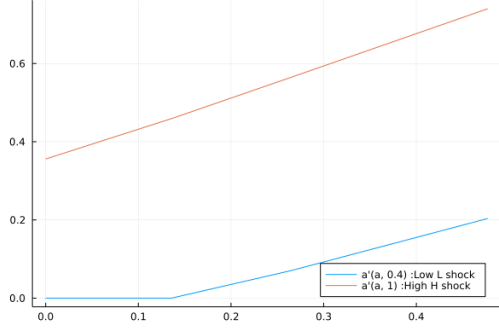


Figure 4: Asset Policy function near zero assets

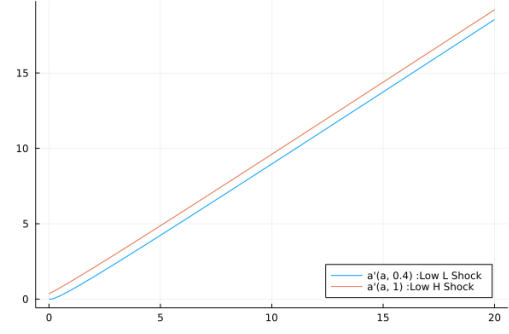


Figure 5: Asset policy function

2.1.1 Discussion of Results

We note that the asset constraint is binding for the low productivity agent ($l = 0.4$) at very low level of asset. At high level of assets the asset constraint is no longer binding and the curve is upward sloping. For the high productivity agent ($l = 1$), the asset policy function is upward sloping at all assets with a positive intercept.