

# Projection methods; TI field course 2010

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## 1 Idea

The idea of this assignment is to show you that solving a model with projection methods is actually quite simple. There is one somewhat tricky step, namely the numerical integration of the conditional expectation.

## 2 Model

The standard growth model characterized by the following equations

$$\begin{aligned}c_t^{-\nu} &= E_t [\beta c_{t+1}^{-\nu} (\alpha \exp(z_{t+1}) k_t^{\alpha-1} + 1 - \delta)] \\c_t + k_t &= \exp(z_t) k_{t-1}^{\alpha} + (1 - \delta) k_{t-1} \\z_{t+1} &= \rho z_t + \sigma \varepsilon_{t+1} \quad \varepsilon_{t+1} \sim N(0, 1)\end{aligned}$$

Of course you can choose your own parameters, but to make programs comparable let's consider

case	B
$\alpha$	0.36
$\rho$	0.95
$\beta$	0.99
$\delta$	0.10
$\nu$	5
$\sigma$	0.005

### 3 Preliminaries

#### Grid

- First calculate the steady state for capital,  $\bar{k}$ .
- For  $k$  take values in the range from  $0.5\bar{k}$  to  $1.5\bar{k}$ .
- For  $z$  construct a grid going from 3 times the standard deviation of  $z$  below the mean of  $z$  to 3 times above. Note that the standard deviation of  $z$  is not equal to the standard deviation of  $\varepsilon$ .

**Approximating function** You always have to take a stand on which function you are going to approximate with what kind of approximating function. Here we approximate the consumption function and use the following second-order polynomial

$$\ln c = \begin{aligned} &\gamma_1 + \gamma_2 \ln k_{-1} + \gamma_3 z \\ &+ \gamma_4 (\ln k_{-1})^2 + \gamma_5 z \ln k_{-1} + \gamma_6 z^2 \end{aligned}$$

The matlab function `consfun.m` defines this function

### 4 To do

1. Write a Matlab function that calculates the sum of squared Euler equation errors on the grid taking as variable input the values of the six  $\gamma$  coefficients
2. Run the Matlab program that finds the values of  $\gamma$  that minimizes the sum of the squared Euler equation errors. As initial values for  $\gamma$  you can use

$$[-0.6889 \ 0.4829 \ 0.7228 \ 0.0297 \ 0.0834 \ -0.0933]$$

These are the values that solve the model when  $\nu = 1$ .

### 5 Numerical integration

Evaluation of the error on the grid requires evaluation of the conditional expectation, i.e., numerical integration. Don't be scared of this. Just realize the following:

- The stochastic variable inside the conditional expectation is  $\varepsilon_{+1}$ . Note that  $\varepsilon_{+1}$  shows up *three* times, because  $z_{+1}$  shows up three times. Once directly, once as an argument of next period's consumption choice,  $c_{+1}$ , and once as an argument of next period's capital choice,  $k_{+1}$ .
- Numerical integration basically comes down to thinking of  $\varepsilon_{+1}$  as a random variable with discrete support with each realization having a probability. That is, in the slides we showed that

$$\mathbb{E} [h (\varepsilon_{+1})],$$

with  $\varepsilon_{+1} \sim N(\mu, \sigma^2)$ , can be approximated as follows:

$$\begin{aligned} \mathbb{E} [h (\varepsilon_{+1})] &= \int_{-\infty}^{\infty} h(\varepsilon_{+1}) \frac{\exp \{ -(\varepsilon_{+1} - \mu)^2 / (2\sigma^2) \}}{\sigma \sqrt{2\pi}} d\varepsilon_{+1} \\ &\approx \sum_{j=1}^J h(\mu + \sigma \sqrt{2}\zeta_j) \frac{\omega_j}{\sqrt{\pi}} \end{aligned}$$

where  $\omega_j$  and  $\zeta_j$  are the Hermite weights (sort of like probabilities) and nodes, respectively. Note that in our application  $\mu = 0$ .