Exercise 1:

(a) $log(z_t)$ follows a stationary AR(1)-process, since $0 < \rho < 1$. Hence, $log(z_t)$ has a time independent expectation and variance.

Erwartungswert:
$$E[log(z_t)] = \rho \underbrace{E[log(z_{t-1})]}_{=E[log(z_t)]} + \underbrace{E[\varepsilon_t]}_{=0}$$

 $\Leftrightarrow E[log(z_t)] = 0$
 $\Leftrightarrow log(E[z_t]) = 0$
 $\Leftrightarrow E(z_t) = 1 = z$

Varianz:
$$var(z_t) = \rho^2 \underbrace{var(z_{t-1})}_{=var(z_t)} + \underbrace{var(\varepsilon_t)}_{=\sigma_{\varepsilon}^2}$$

$$var(z_t) = \frac{\sigma_{\varepsilon}^2}{1 - \rho^2}$$

(b) States k_t, z_t , control c_t . Bellmann-Equation:

$$V(k_t) = \max_{k_{t+1}} \left\{ U(\underbrace{c(k_t, k_{t+1})}_{c_t}) + \beta E_t V(k_{t+1}) \right\}$$
 Nebenbed: $c_t = f(k_t) + (1 - \delta)k_t - k_{t+1}$

Consider:

$$\frac{\partial V(k_t)}{\partial k_t} = U'(c_t) \frac{\partial c_t}{\partial k_t} = U'(c_t) (f'(k_t) + (1 - \delta))$$

$$\Rightarrow \frac{\partial V(k_{t+1})}{\partial k_{t+1}} = U'(c_{t+1}) (f'(k_{t+1}) + (1 - \delta))$$

Thus, we get:

$$\begin{split} \frac{\partial V(k_t)}{\partial k_{t+1}} &= U'(c_t) \frac{\partial c_t}{\partial k_{t+1}} + \beta E_t \frac{\partial V(k_{t+1})}{\partial k_{t+1}} \\ &= U'(c_t)(-1) + \beta E_t \left[U'(c_{t+1})(f'(k_{t+1}) + (1 - \delta)) \right] \\ \Leftrightarrow c_t^{-\sigma} &= \beta E_t \left[c_{t+1}^{-\sigma} (\alpha z_{t+1} k_{t+1}^{\alpha - 1} + (1 - \delta)) \right] \end{split}$$

This is the Euler-equation.

(c) Set up the Lagrangian:

$$L = E_t \sum_{t=0}^{\infty} \beta^{t+s} \left[\frac{c_{t+s}^{1-\sigma} - 1}{1-\sigma} - \lambda_{t+s} \left(c_{t+s} + k_{t+s+1} - z_{t+s} k_{t+s}^{\alpha} - (1-\delta) k_{t+s} \right) \right]$$

$$\frac{\partial L}{\partial c_t} = E_t \left[\beta^t \left(c_t^{-\sigma} - \lambda_t \right) \right] = 0 \Leftrightarrow \lambda_t = c_t^{-\sigma}$$

$$\frac{\partial L}{\partial k_{t+1}} = E_t \left[\beta^t (-\lambda_t) + \beta^{t+1} (-\lambda_{t+1}) (-\alpha z_{t+1} k_{t+1}^{\alpha-1} - 1 + \delta) \right] = 0$$

$$\Leftrightarrow \lambda_t = E_t \beta \lambda_{t+1} (\alpha z_{t+1} k_{t+1}^{\alpha-1} + 1 - \delta)$$

$$\Rightarrow c_t^{-\sigma} = \beta E_t \left[c_{t+1}^{-\sigma} (\alpha z_{t+1} k_{t+1}^{\alpha-1} + (1-\delta)) \right]$$

The same Euler-equation.

(d) In steady-state there is no uncertainty and all time-subscripts cancel: $c_t = c_{t+1} = c, k_t = c_{t+1}$ $k_{t+1} = k, z = E(z_t) = 1$. From the Euler equation you get:

$$c^{-\sigma} = \beta c^{-\sigma} (\alpha k^{\alpha - 1} + 1 - \delta)$$

$$\frac{1}{\beta} = \alpha k^{\alpha - 1} + 1 - \delta$$

$$\Leftrightarrow \alpha k^{\alpha - 1} = \frac{1}{\beta} - (1 - \delta)$$

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

Linearization

Linearization of the Euler-equation:

From steady-state you know: $\frac{1}{\beta} = \alpha z k^{\alpha-1} + 1 - \delta \Leftrightarrow -\log(\beta) = \log(\alpha z k^{\alpha-1} + 1 - \delta)$ There are 2 methods to do this:

(1) Take logs and use Taylor-Approximation:

$$c_t^{-\sigma} = \beta E_t \left[c_{t+1}^{-\sigma} (\alpha z_{t+1} k_{t+1}^{\alpha - 1} + 1 - \delta) \right]$$

$$-\sigma log(c_t) = \log(\beta) - \sigma log(E_t c_{t+1}) + log(\alpha E_t z_{t+1} E_t k_{t+1}^{\alpha - 1} + 1 - \delta)$$

$$\sigma \left[log(E_t c_{t+1}) - log(c_t) \right] = \log(\beta) + log(\alpha E_t z_{t+1} E_t k_{t+1}^{\alpha - 1} + 1 - \delta)$$

Approximate the left side:

$$\sigma \left[\log(c) - \log(c) \right] + \frac{\sigma}{c} \left(E_t c_{t+1} - c \right) - \frac{\sigma}{c} \left(c_t - c \right) = \sigma \left(E_t \widehat{c}_{t+1} - \widehat{c}_t \right)$$

Approximate the right side:

$$\log(\beta) + \underbrace{\log(\alpha z k^{\alpha - 1} + 1 - \delta)}_{-\log(\beta)}$$

$$+ \underbrace{\frac{\alpha k^{\alpha - 1}}{\alpha z k^{\alpha - 1} + 1 - \delta}}_{=1/\beta} (E_t z_{t+1} - z) \underbrace{\frac{1}{z}}_{=1}$$

$$+ \beta \alpha (\alpha - 1) z k^{\alpha - 2} (E_t k_{t+1} - k)$$

$$+ \beta \alpha (\alpha - 1) k^{\alpha - 1} E_t \widehat{k}_{t+1}$$

Thus, the approximation yields:

$$\sigma\left(E_t \widehat{c}_{t+1} - \widehat{c}_t\right) = \beta \alpha k^{\alpha - 1} E_t \widehat{z}_{t+1} + \beta \alpha (\alpha - 1) k^{\alpha - 1} E_t \widehat{k}_{t+1}$$

(2) Rewrite the model using $x_t = e^{\log(x_t) - \log(x) + \log(x)} = xe^{\log(x_t) - \log(x)} \approx xe^{\hat{x}_t}$ and approximate for \hat{x}_t :

$$c^{-\sigma}e^{-\sigma\widehat{c}_t} = \beta \left[c^{-\sigma}e^{-\sigma E_t\widehat{c}_{t+1}} \left(\alpha z k^{\alpha - 1} e^{E_t\widehat{z}_{t+1} + (\alpha - 1)E_t\widehat{k}_{t+1}} + 1 - \delta \right) \right]$$
$$e^{\sigma E_t\widehat{c}_{t+1} - \sigma\widehat{c}_t} = \beta \left(\alpha z k^{\alpha - 1} e^{E_t\widehat{z}_{t+1} + (\alpha - 1)E_t\widehat{k}_{t+1}} + 1 - \delta \right)$$

Approximation of the LHS:

$$e^{0} + e^{0}\sigma(E_{t}\widehat{c}_{t+1} - 0) - e^{0}\sigma(\widehat{c}_{t} - 0) = 1 + \sigma(E_{t}\widehat{c}_{t+1} - \widehat{c}_{t})$$

Approximation of the RHS:

$$\underbrace{\frac{\beta \left(\alpha z k^{\alpha-1} e^0 + 1 - \delta\right)}{=1}}_{=1}$$
$$+\beta \alpha z k^{\alpha-1} e^0 (E_t \widehat{z}_{t+1} - 0)$$
$$+\beta \alpha z k^{\alpha-1} e^0 (\alpha - 1) (E_t \widehat{k}_{t+1} - 0)$$

This is the same linearized Euler-equation.

For the capital-accumulation-equation you get:

$$\widehat{k}_{t+1} = k^{\alpha - 1} \widehat{z}_t - \frac{c}{k} \widehat{c}_t + \frac{1}{\beta} \widehat{k}_t$$

And the technological process (just take logs and subtract the steady-state):

$$\widehat{z}_t = \rho z_{t-1} + \varepsilon_t$$

(e) Put all Variables with time-subscript t+1 (except shocks) on the left side:

$$-1E_{t}\widehat{c}_{t+1} - \frac{\beta\alpha(\alpha - 1)k^{\alpha - 1}}{\sigma}E_{t}\widehat{k}_{t+1} - \frac{\beta\alpha k^{\alpha - 1}}{\sigma}E_{t}\widehat{z}_{t+1} = \widehat{c}_{t}$$

$$\widehat{k}_{t+1} = -\frac{c}{k}\widehat{c}_{t} + \frac{1}{\beta}\widehat{k}_{t} + k^{\alpha - 1}\widehat{z}_{t}$$

$$\widehat{z}_{t+1} = \rho z_{t} + \varepsilon_{t+1}$$

Thus,
$$\mathbf{x_{t+1}} = \begin{pmatrix} c_{t+1} \\ k_{t+1} \\ z_{t+1} \end{pmatrix}$$
, $\boldsymbol{\nu_{t+1}} = (\varepsilon_{t+1})$, $\boldsymbol{\eta_{t+1}} = \begin{pmatrix} E_t c_{t+1} - c_{t+1} \\ E_t k_{t+1} - k_{t+1} \\ E_t z_{t+1} - z_{t+1} \end{pmatrix}$. The system can then be

written as:

$$\underbrace{\begin{pmatrix} -1 & -\frac{\beta\alpha(\alpha-1)k^{\alpha-1}}{\sigma} & -\frac{\beta\alpha k^{\alpha-1}}{\sigma} \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}}_{\mathbf{A}} \underbrace{\begin{pmatrix} c_{t+1} \\ k_{t+1} \\ z_{t+1} \end{pmatrix}}_{\mathbf{x}_{t+1}} = \underbrace{\begin{pmatrix} 1 & 0 & 0 \\ -\frac{c}{k} & \frac{1}{\beta} & k^{\alpha-1} \\ 0 & 0 & \rho \end{pmatrix}}_{\mathbf{x}_{t}} \underbrace{\begin{pmatrix} c_{t} \\ k_{t} \\ z_{t} \end{pmatrix}}_{\mathbf{x}_{t}} + \underbrace{\begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}}_{\boldsymbol{\nu}_{t+1}} \underbrace{\begin{pmatrix} 1 & \frac{\beta\alpha(\alpha-1)k^{\alpha-1}}{\sigma} & \frac{\beta\alpha k^{\alpha-1}}{\sigma} \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}}_{\boldsymbol{\eta}_{t,t+1}} \underbrace{\begin{pmatrix} \eta_{1,t+1} \\ \eta_{2,t+1} \\ \eta_{3,t+1} \end{pmatrix}}_{\boldsymbol{\eta}_{t+1}}$$

Vectorization and Kroneckerproduct

Example for vectorization:

$$vec\begin{pmatrix} 1 & 3 & 2 \\ 1 & 0 & 0 \\ 1 & 2 & 2 \end{pmatrix} = \begin{pmatrix} 1 & 1 & 1 & 3 & 0 & 2 & 2 & 0 & 2 \end{pmatrix}'$$

Example for Kroneckerproduct:

$$\underbrace{\begin{pmatrix} 1 & 3 & 2 \\ 1 & 0 & 0 \\ 1 & 2 & 2 \end{pmatrix}}_{3 \times 3} \otimes \underbrace{\begin{pmatrix} 0 & 5 \\ 5 & 0 \\ 1 & 1 \end{pmatrix}}_{3 \times 2} = \underbrace{\begin{pmatrix} 1 \cdot \begin{pmatrix} 0 & 5 \\ 5 & 0 \\ 1 & 1 \end{pmatrix} & 3 \cdot \begin{pmatrix} 0 & 5 \\ 5 & 0 \\ 1 & 1 \end{pmatrix} & 2 \cdot \begin{pmatrix} 0 & 5 \\ 5 & 0 \\ 1 & 1 \end{pmatrix} \\ 0 & 5 \\ 1 \cdot \begin{pmatrix} 0 & 5 \\ 5 & 0 \\ 1 & 1 \end{pmatrix} & 0 \cdot \begin{pmatrix} 0 & 5 \\ 5 & 0 \\ 1 & 1 \end{pmatrix} \\ 0 & 5 \\ 1 \cdot \begin{pmatrix} 0 & 5 \\ 5 & 0 \\ 1 & 1 \end{pmatrix} & 2 \cdot \begin{pmatrix} 0 & 5 \\ 5 & 0 \\ 1 & 1 \end{pmatrix} \\ 0 & 5 \\ 1 \cdot \begin{pmatrix} 0 & 5 \\ 5 & 0 \\ 1 & 1 \end{pmatrix} & 2 \cdot \begin{pmatrix} 0 & 5 \\ 5 & 0 \\ 1 & 1 \end{pmatrix} \\ 0 & 5 \\ 5 & 0 \\ 1 & 1 \end{pmatrix} & 2 \cdot \begin{pmatrix} 0 & 5 \\ 5 & 0 \\ 1 & 1 \end{pmatrix}$$

Consider the matrices A: $m \times n$, B: $n \times p$ and C: $p \times k$. Show that $vec(ABC) = (C' \otimes A) vec(B)$.

$$ABC = A \begin{pmatrix} b_1 & b_2 & \dots & b_p \end{pmatrix} \begin{pmatrix} c_{11} & c_{12} & \dots & c_{1k} \\ c_{21} & c_{22} & \dots & c_{2k} \\ \vdots & \vdots & \vdots & \vdots \\ c_{p1} & c_{p2} & \dots & c_{pk} \end{pmatrix}$$

$$= A \underbrace{\begin{pmatrix} b_1 c_{11} + b_2 c_{21} + \dots + b_p c_{p1}, & b_1 c_{12} + b_2 c_{22} + \dots + b_p c_{p2}, & \dots, & b_1 c_{1k} + b_2 c_{2k} + \dots + b_p c_{pk} \end{pmatrix}}_{n \times k}$$

$$vec(ABC) = \begin{pmatrix} c_{11}Ab_1 + c_{21}Ab_2 + \dots + c_{p1}Ab_p \\ c_{12}Ab_1 + c_{22}Ab_2 + \dots + c_{p2}Ab_p \\ \vdots \\ c_{1k}Ab_1 + c_{2k}Ab_2 + \dots + c_{pk}Ab_p \end{pmatrix} = \begin{pmatrix} c_{11}A & c_{21}A & \dots & c_{p1}A \\ c_{12}A & c_{22}A & \dots & c_{p2}A \\ \vdots & \vdots & \vdots & \vdots \\ c_{1k}A & c_{2k}A & \dots & c_{pk}A \end{pmatrix} \begin{pmatrix} b_1 \\ b_2 \\ \vdots \\ b_p \end{pmatrix} = (C' \otimes A) vec(B)$$

Exercise 5

Derivation of optimality-conditions:

•
$$u_c(c_t, l_t) = \theta \left[c_t^{\theta} (1 - l_t)^{1-\theta} \right]^{1-\tau} c_t^{-1}$$

•
$$u_l(c_t, l_t) = -(1 - \theta) \left[c_t^{\theta} (1 - l_t)^{1 - \theta} \right]^{1 - \tau} (1 - l_t)^{-1}$$

•
$$f_k = \alpha \left[\alpha k_{t-1}^{\psi} + (1 - \alpha) l_t^{\psi} \right]^{\frac{1 - \psi}{\psi}} k_{t-1}^{\psi - 1} = \alpha \left(\frac{f_t}{k_{t-1}} \right)^{1 - \psi} = \alpha \left[\alpha + (1 - \alpha) \left(\frac{l_t}{k_{t-1}} \right)^{\psi} \right]^{\frac{1 - \psi}{\psi}}$$

•
$$f_l = (1-\alpha) \left[\alpha k_{t-1}^{\psi} + (1-\alpha) l_t^{\psi} \right]^{\frac{1-\psi}{\psi}} l_t^{\psi-1} = (1-\alpha) \left(\frac{f_t}{l_t} \right)^{1-\psi} = (1-\alpha) \left[\alpha \left(\frac{k_{t-1}}{l_t} \right)^{\psi} + (1-\alpha) \right]^{\frac{1-\psi}{\psi}}$$

• Euler:

$$\left[c_t^{\theta} (1 - l_t)^{1-\theta} \right]^{1-\tau} c_t^{-1} = \beta E_t \left[c_{t+1}^{\theta} (1 - l_{t+1})^{1-\theta} \right]^{1-\tau} c_{t+1}^{-1} \cdot \left[A_{t+1} \alpha \left(\frac{f_{t+1}}{k_t} \right)^{1-\psi} + 1 - \delta \right]$$

$$= \beta E_t \left[c_{t+1}^{\theta} (1 - l_{t+1})^{1-\theta} \right]^{1-\tau} c_{t+1}^{-1} \cdot \left[\alpha A_{t+1}^{\psi} \left(\frac{y_{t+1}}{k_t} \right)^{1-\psi} + 1 - \delta \right]$$

• Consumption-Leisure:

$$\frac{1-\theta}{\theta}\frac{c_t}{1-l_t} = A_t(1-\alpha)\left(\frac{f_t}{l_t}\right)^{1-\psi} = A_t^{\psi}(1-\alpha)\left(\frac{y_t}{l_t}\right)^{1-\psi}$$

• Resource-constraint:

$$y_t = c_t + k_t - (1 - \delta)k_{t-1}$$

• Stochastic process: $A_t = e^{a_t}$ and $a_t = \rho a_{t-1} + \sigma \varepsilon_t$ with $\varepsilon_t \sim N(0, 1)$.

Derivation of steady-state:

- 1. $A = A^{ss}$
- 2. From Euler:

$$\begin{split} &\frac{1}{\beta} = A^{\psi} \alpha \left(\frac{y}{k}\right)^{1-\psi} + 1 - \delta \\ &\Leftrightarrow \frac{y}{k} = \left(\frac{\beta^{-1} - 1 + \delta}{\alpha A^{\psi}}\right)^{\frac{1}{1-\psi}} \end{split}$$

3. From resource-constraint:

$$\frac{c}{k} = \frac{y}{k} - \delta$$

4. Definition of Production-function:

$$\frac{y}{k} = A \left[\alpha + (1 - \alpha) \left(\frac{l}{k} \right)^{\psi} \right]^{\frac{1}{\psi}} = \left(\frac{\beta^{-1} - 1 + \delta}{\alpha A^{\psi}} \right)^{\frac{1}{1 - \psi}} = \frac{y}{k}$$

$$\Leftrightarrow \frac{l}{k} = \left[\left(\left(\frac{y/k}{A} \right)^{\psi} - \alpha \right) (1 - \alpha)^{-1} \right]^{\frac{1}{\psi}}$$

5. Identity:

$$\frac{y}{l} = \frac{y}{k} \frac{k}{l}$$

6. Identity:

$$\frac{c}{l} = \frac{c}{k} \frac{k}{l}$$

7. From the consumption-leisure decision:

$$\begin{split} l\frac{c}{l} &= (1 - l)\frac{\theta}{1 - \theta}A^{\psi}(1 - \alpha)\left(\frac{y}{l}\right)^{1 - \psi} \\ \Leftrightarrow l &= \left(1 + \frac{\frac{c}{l}}{\frac{\theta(1 - \alpha)}{1 - \theta}A^{\psi}\left(\frac{y}{l}\right)^{1 - \psi}}\right)^{-1} \end{split}$$

8. Identities:

$$c = \frac{c}{l}l, \qquad k = \frac{l}{l/k}, \qquad y = \frac{y}{k}k.$$

How to calibrate?

 \bullet We have a CES-production-function. One can show that the capital-share in steady-state ist equal to

$$s(l,k) = \frac{\alpha k^{\psi}}{\alpha k^{\psi} + (1-\alpha)l^{\psi}} = \frac{\alpha k^{\psi}}{\left[\alpha + (1-\alpha)\left(\frac{l}{k}\right)^{\psi}\right]k^{\psi}} = \frac{\alpha}{\left(\frac{y}{Ak}\right)^{\psi}}$$
$$= \frac{\alpha A^{\psi}}{\left(\frac{\beta^{-1}-1+\delta}{\alpha}\right)^{\frac{\psi}{1-\psi}}} = \frac{\alpha^{\frac{1}{1-\psi}}A^{\psi}}{(\beta^{-1}-1+\delta)^{\frac{\psi}{1-\psi}}}$$

• So, choose parameters such that this expression is close to 0.2-0.3.