

The Classical Monetary Model

(Stochastic Version; From “Economic Growth: Theory and Numerical Solution Methods”, by A. Novales, E, Fernández, y J. Ruiz, 2009, Springer Verlag)

The representative agent solves the problem:

$$\underset{\{c_t, M_{t+1}, k_{t+1}, V_{t+1}\}_{t=0}^{\infty}}{\text{Max}} E_0 \sum_{t=0}^{\infty} \beta^t U(c_t, M_{t+1}/P_t)$$

subject to:

$$(1 + \tau^c)c_t + k_{t+1} - (1 - \delta)k_t + \frac{M_{t+1}}{P_t} + \left[\frac{V_{t+1}}{(1 + i_t)P_t} \right] = (1 - \tau^y)A_t k_t^\alpha + \frac{M_t}{P_t} + \frac{V_t}{P_t} + \gamma_t,$$

given k_0, M_0, V_0 . The productivity shock is assumed to obey the stochastic process

$$\ln A_t = (1 - \rho_A) \ln A_{ss} + \rho_A \ln A_{t-1} + \varepsilon_{A,t}, \quad |\rho_A| < 1, \quad \varepsilon_{A,t} \underset{iid}{\sim} N(0, \sigma_A^2).$$

The government raises income and consumption taxes, prints money and issues nominal bonds, which are bought at price $1/(1+i_t)$. It also provides transfers γ_t to the private sector. The government budget constraint is,

$$\tau^c c_t + \tau^y A_t k_t^\alpha + \frac{M_{t+1} - M_t}{P_t} + \left[\frac{V_{t+1}}{(1+i_t)P_t} - \frac{V_t}{P_t} \right] = \gamma_t.$$

The fiscal authority chooses sequences $\{\tau^c, \tau^y\}$ and $\{\gamma_t\}_{t=0}^\infty$. To guarantee stability of the public debt trajectory, we will assume that:

$$\gamma_t = \bar{\gamma} - \eta \frac{V_t}{P_t}, \quad (6.147)$$

which implies,

$$\tilde{b}_{t+1} = (1+i_t) \left[\gamma - \tau^c c_t - \tau^y A_t k_t^\alpha - \tilde{m}_{t+1} + \frac{\tilde{m}_t}{1+\pi_t} \right] + (1-\eta) \frac{1+i_t}{1+\pi_t} \tilde{b}_t. \quad (6.148)$$

where, as in previous sections, $\tilde{b}_{t+1} = \frac{V_{t+1}}{P_t}$, $\tilde{m}_{t+1} = \frac{M_{t+1}}{P_t}$. Then, \tilde{b}_{t+1} will be stable so long as $1 > \eta > \frac{r_{ss}}{1+r_{ss}}$, where: $1+r_{ss} = (1+i_{ss})/(1+\pi_{ss})$.

Additionally, we will assume that the monetary authority chooses the sequence of nominal interest rates $\{i_t\}_{t=0}^{\infty}$ according to a Taylor's rule:

$$\hat{i}_t = \rho_i \hat{i}_{t-1} + \rho_{\pi} \hat{\pi}_t + \rho_y \hat{y}_t + \varepsilon_{i,t}, \quad |\rho_i| < 1, \quad \varepsilon_{i,t} \underset{iid}{\sim} N(0, \sigma_i^2),$$

$$\text{where } \hat{i}_t \equiv \ln \left(\frac{1+i_t}{1+i_{ss}} \right); \quad \hat{\pi}_t \equiv \ln \left(\frac{1+\pi_t}{1+\pi_{ss}} \right); \quad \hat{y}_t \equiv \ln \left(\frac{y_t}{y_{ss}} \right) \underset{y_t = A_t k_t^\alpha}{=} \hat{A}_t + \alpha \hat{k}_t,$$

Case 1: $\rho_{\pi} = 0$.

If $\rho_{\pi} = 0$, the Taylor rule becomes:

$$\hat{i}_t = \rho_i \hat{i}_{t-1} + \rho_y \left(\hat{A}_t + \alpha \hat{k}_t \right) + \varepsilon_{i,t}, \quad (6.149)$$

The Lagrangian of the representative agent is,

$$L = E_0 \left[\sum_{t=0}^{\infty} \beta^t \left\{ U(c_t, M_{t+1}/P_t) + \lambda_t \left(\begin{array}{l} (1 - \tau^y)A_t k_t^\alpha + \frac{M_t}{P_t} + \frac{V_t}{P_t} + \gamma_t - (1 + \tau^c)c_t - k_{t+1} + \\ (1 - \delta)k_t - \frac{M_{t+1}}{P_t} - \left[\frac{V_t}{(1+i_t)P_t} \right] \end{array} \right) \right\} \right]$$

with first order conditions:

$$\begin{aligned} U_c(c_t, \tilde{m}_{t+1}) &= (1 + \tau^c)\lambda_t, \\ \lambda_t &= \beta E_t \left[\lambda_{t+1} \left((1 - \tau^y)\alpha A_{t+1} k_{t+1}^{\alpha-1} + 1 - \delta \right) \right], \\ -U_m(c_t, \tilde{m}_{t+1}) \frac{1}{P_t} + \lambda_t \frac{1}{P_t} &= \beta E_t \left(\lambda_{t+1} \frac{1}{P_{t+1}} \right), \\ \lambda_t \frac{1}{P_t(1+i_t)} &= \beta E_t \left(\lambda_{t+1} \frac{1}{P_{t+1}} \right), \end{aligned}$$

where $\tilde{m}_{t+1} = M_{t+1}/P_t$.

Plugging the first optimality condition into the other three conditions, we get:

$$U_c(c_t, \tilde{m}_{t+1}) = \beta E_t \left[U_c(c_{t+1}, \tilde{m}_{t+2}) \left((1 - \tau^y) \alpha A_{t+1} k_{t+1}^{\alpha-1} + 1 - \delta \right) \right]. \quad (6.150)$$

$$-U_m(c_t, \tilde{m}_{t+1}) + \frac{U_c(c_t, \tilde{m}_{t+1})}{(1 + \tau^e)} = \beta E_t \left[\frac{U_c(c_{t+1}, \tilde{m}_{t+2})}{(1 + \tau^e)} \frac{1}{1 + \pi_{t+1}} \right]. \quad (6.151)$$

$$U_c(c_t, \tilde{m}_{t+1}) = \beta(1 + i_t) E_t \left[U_c(c_{t+1}, \tilde{m}_{t+2}) \frac{1}{1 + \pi_{t+1}} \right]. \quad (6.152)$$

From (6.151) and (6.152) we get:

$$\frac{i_t}{1+i_t} = (1 + \tau^c) \frac{U_m(c_t, \tilde{m}_{t+1})}{U_c(c_t, \tilde{m}_{t+1})}. \quad (6.153)$$

For the utility function $U(c_t, M_{t+1}/P_t) = \frac{[c_t(M_{t+1}/P_t)^\theta]^{1-\sigma} - 1}{1-\sigma}$, $\sigma > 0$, (6.153) becomes:

$$\frac{i_t}{1+i_t} = \theta(1 + \tau^c) \frac{c_t}{\tilde{m}_{t+1}}. \quad (6.154)$$

From the budget constraints for the representative agent and the government, we obtain the global constraint of resources in the economy:

$$c_t + k_{t+1} - (1 - \delta)k_t = A_t k_t^\alpha. \quad (6.155)$$

Let us now recollect the set of optimality conditions that we are going to log-linearize in order to solve the model under the assumed utility function. From (6.150), (6.152), (6.154), (6.155), we get,

$$c_t^{-\sigma} \tilde{m}_{t+1}^{\theta(1-\sigma)} = \beta E_t \left[c_{t+1}^{-\sigma} \tilde{m}_{t+2}^{\theta(1-\sigma)} \left((1 - \tau^y) \alpha A_{t+1} k_{t+1}^{\alpha-1} + 1 - \delta \right) \right], \quad (6.156)$$

$$c_t^{-\sigma} \tilde{m}_{t+1}^{\theta(1-\sigma)} = \beta(1 + i_t) E_t \left[c_{t+1}^{-\sigma} \tilde{m}_{t+2}^{\theta(1-\sigma)} \frac{1}{1 + \pi_{t+1}} \right], \quad (6.157)$$

$$\theta(1 + \tau^c) \frac{c_t}{\tilde{m}_{t+1}} + \frac{1}{1 + i_t} - 1 = 0, \quad (6.158)$$

$$c_t + k_{t+1} - (1 - \delta)k_t = A_t k_t^\alpha. \quad (6.159)$$

Steady-state

From (6.156), (6.157), (6.158), (6.159), we get:

$$k_{ss} = \left[\frac{(1 - \tau^y)\alpha A_{ss}}{\frac{1}{\beta} - (1 - \delta)} \right]^{\frac{1}{1-\alpha}}.$$

$$c_{ss} = A_{ss}k_{ss}^\alpha - \delta k_{ss}.$$

$$\pi_{ss} = \beta(1 + i_{ss}) - 1.$$

$$m_{ss} = \theta(1 + \tau^c)c_{ss} \frac{1 + i_{ss}}{i_{ss}}.$$

From (6.148) we get b_{ss} , while γ_{ss} is obtained from (6.147).

Log-linearization

To log-linearize this system (6.156)-(6.159), we rewrite the system as,

$$e^{-\sigma(\ln c_t)} e^{\theta(1-\sigma)(\ln \tilde{m}_{t+1})} = \beta E_t \left[e^{-\sigma(\ln c_{t+1})} e^{\theta(1-\sigma)(\ln \tilde{m}_{t+2})} \left((1 - \tau^y) \alpha e^{\ln A_{t+1}} e^{(\alpha-1)(\ln k_{t+1})} + 1 - \delta \right) \right], \quad (6.160)$$

$$e^{-\sigma(\ln c_t)} e^{\theta(1-\sigma)(\ln \tilde{m}_{t+1})} = \beta e^{\tilde{i}_t} E_t \left[e^{-\sigma(\ln c_{t+1})} e^{\theta(1-\sigma)(\ln \tilde{m}_{t+2})} e^{-\ln \pi_{t+1}} \right], \quad (6.161)$$

$$\theta(1 + \tau^c) e^{\ln c_t - \ln \tilde{m}_{t+1}} + e^{-\ln(1+i_t)} - 1 = 0, \quad (6.162)$$

$$e^{\ln c_t} + e^{\ln k_{t+1}} - (1 - \delta) e^{\ln k_t} = e^{\tilde{A}_t} e^{\alpha(\ln k_t)}, \quad (6.163)$$

Let us denote the deviations from steady-state by: $\hat{u}_t = \ln(u_t/u_{ss})$, $u = c, k, \tilde{m}, A, 1 + i, 1 + \pi$. Using the fact that $E_t \hat{A}_{t+1} = \rho_A \hat{A}_t$, and also that $(1 - \tau^y) \alpha A_{ss} k_{ss}^{\alpha-1} = \frac{1}{\beta} - (1 - \delta)$, we obtain from (6.160)-(6.163),

$$0 = \frac{\sigma}{\beta} \hat{c}_t - \frac{\theta(1-\sigma)}{\beta} \hat{m}_{t+1} - \frac{\sigma}{\beta} E_t \hat{c}_{t+1} + \frac{\theta(1-\sigma)}{\beta} E_t \hat{m}_{t+2} + \left(\frac{1}{\beta} - (1-\delta) \right) \rho_A \hat{A}_t + \left(\frac{1}{\beta} - (1-\delta) \right) (\alpha - 1) \hat{k}_{t+1}. \quad (6.164)$$

$$0 = \frac{\sigma}{\beta} \hat{c}_t - \frac{\theta(1-\sigma)}{\beta} \hat{m}_{t+1} + \frac{1+i_{ss}}{1+\pi_{ss}} \hat{i}_t - \frac{\sigma}{\beta} E_t \hat{c}_{t+1} + \frac{\theta(1-\sigma)}{\beta} E_t \hat{m}_{t+2} - \frac{1}{\beta} E_t \hat{\pi}_{t+1}. \quad (6.165)$$

$$\hat{m}_{t+1} = \hat{c}_t - \frac{1}{i_{ss}} \hat{i}_t \quad (6.166)$$

$$A_{ss} k_{ss}^\alpha \hat{A}_t + (A_{ss} k_{ss}^\alpha + (1-\delta) k_{ss}) \hat{k}_t - c_{ss} \hat{c}_t - k_{ss} \hat{k}_{t+1} = 0 \quad (6.167)$$

where \hat{m}_{t+1} is known at time t .

Plugging (6.166) into (6.164) and (6.165), we get:

$$0 = \frac{\sigma - \theta(1 - \sigma)}{\beta} \hat{c}_t + \frac{\theta(1 - \sigma)}{\beta i_{ss}} \hat{i}_t - \frac{\sigma - \theta(1 - \sigma)}{\beta} E_t \hat{c}_{t+1} - \frac{\theta(1 - \sigma)}{\beta i_{ss}} E_t \hat{i}_{t+1} + \left(\frac{1}{\beta} - (1 - \delta) \right) \rho_A \hat{A}_t + \left(\frac{1}{\beta} - (1 - \delta) \right) (\alpha - 1) \hat{k}_{t+1}, \quad (6.168)$$

$$0 = \frac{\sigma - \theta(1 - \sigma)}{\beta} \hat{c}_t + \left(\frac{\theta(1 - \sigma)}{\beta i_{ss}} + \frac{1 + i_{ss}}{1 + \pi_{ss}} \right) \hat{i}_t - \frac{\sigma - \theta(1 - \sigma)}{\beta} E_t \hat{c}_{t+1} - \frac{\theta(1 - \sigma)}{\beta i_{ss}} E_t \hat{i}_{t+1} - \frac{1}{\beta} E_t \hat{\pi}_{t+1}. \quad (6.169)$$

We will now use the result, shown in Appendix 1, that in a log-linear approximation,

$$E_t \hat{\pi}_{t+1} = \hat{i}_t - \hat{r}_t \quad (6.170)$$

and also that:

$$\hat{r}_t = \frac{1}{1 + r_{ss}} (1 - \tau^y) \alpha A_{ss} k_{ss}^{\alpha-1} \left[\rho_A \hat{A}_t + (\alpha - 1) \hat{k}_{t+1} \right] \quad (6.171)$$

From (6.149):

$$E_t \hat{i}_{t+1} = \rho_i \hat{i}_t + \rho_y \left(\rho_A \hat{A}_t + \alpha \hat{k}_{t+1} \right). \quad (6.172)$$

Using (6.170)-(6.172), we get from (6.168) :

$$\begin{aligned}
0 = & \frac{\sigma - \theta(1 - \sigma)}{\beta} \hat{c}_t + \frac{\theta(1 - \sigma)}{\beta i_{ss}} (1 - \rho_i) \hat{i}_t - \frac{\sigma - \theta(1 - \sigma)}{\beta} E_t \hat{c}_{t+1} + \\
& \left[\left(\frac{1}{\beta} - (1 - \delta) \right) - \frac{\theta(1 - \sigma)}{\beta i_{ss}} \rho_y \right] \rho_A \hat{A}_t - \\
& \left[(1 - \alpha) \left(\frac{1}{\beta} - (1 - \delta) \right) + \frac{\theta(1 - \sigma)}{\beta i_{ss}} \alpha \rho_y \right] \hat{k}_{t+1},
\end{aligned} \tag{6.173}$$

while from (6.169) we get:

$$\begin{aligned}
0 = & \frac{\sigma - \theta(1 - \sigma)}{\beta} \hat{c}_t + \frac{\theta(1 - \sigma)}{\beta i_{ss}} (1 - \rho_i) \hat{i}_t - \frac{\sigma - \theta(1 - \sigma)}{\beta} E_t \hat{c}_{t+1} + \\
& \left[\left(\frac{1}{\beta} - (1 - \delta) \right) - \frac{\theta(1 - \sigma)}{\beta i_{ss}} \rho_y \right] \rho_A \hat{A}_t - \\
& \left[(1 - \alpha) \left(\frac{1}{\beta} - (1 - \delta) \right) + \frac{\theta(1 - \sigma)}{\beta i_{ss}} \alpha \rho_y \right] \hat{k}_{t+1},
\end{aligned} \tag{6.174}$$

Therefore, we can initially reduce the solution to the model to equations (6.163), (6.173) and (6.149), whose matrix representation is:

$$\begin{aligned}
 & \underbrace{\begin{bmatrix} 0 & k_{ss} & 0 \\ \frac{\sigma - \theta(1-\sigma)}{\beta} & \left[(1-\alpha) \left(\frac{1}{\beta} - (1-\delta) \right) + \frac{\theta(1-\sigma)}{\beta i_{ss}} \alpha \rho_y \right] & -\frac{\theta(1-\sigma)}{\beta i_{ss}} (1-\rho_i) \\ 0 & 0 & 1 \end{bmatrix}}_D \underbrace{\begin{bmatrix} E_t \hat{c}_{t+1} \\ \hat{k}_{t+1} \\ \hat{i}_t \end{bmatrix}}_{E_t v_{t+1}} \\
 = & \underbrace{\begin{bmatrix} -c_{ss} & \alpha A_{ss} k_{ss}^\alpha + (1-\delta) k_{ss} & 0 \\ \frac{\sigma - \theta(1-\sigma)}{\beta} & 0 & 0 \\ 0 & \alpha \rho_y & \rho_i \end{bmatrix}}_G \underbrace{\begin{bmatrix} \hat{c}_t \\ \hat{k}_t \\ \hat{i}_{t-1} \end{bmatrix}}_{v_t} + \\
 & \underbrace{\begin{bmatrix} A_{ss} k_{ss}^\alpha \\ \left(\frac{1}{\beta} - (1-\delta) \right) - \frac{\theta(1-\sigma)}{\beta i_{ss}} \rho_y \\ \rho_y \end{bmatrix}}_H \rho_A \hat{A}_t + \underbrace{\begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}}_J \varepsilon_{i,t}
 \end{aligned}$$

that is,

$$DE_t v_{t+1} = Gv_t + H\rho_A \hat{A}_t + J\varepsilon_{i,t},$$

or,

$$E_t v_{t+1} = \Gamma_1 v_t + \Gamma_2 \hat{A}_t + \Gamma_3 \varepsilon_{i,t}, \quad (6.175)$$

where $\Gamma_1 = D^{-1}G$, $\Gamma_2 = D^{-1}H\rho_A$, $\Gamma_3 = D^{-1}J$. Matrix Γ_1 is 3×3 , with two stable and one unstable eigenvalues, whose associated eigenvector allows us to compute the value of the control variable (consumption) as a function of the two states ($\{\hat{k}_t, \hat{i}_{t-1}\}$).

As in previous model economies, we now apply Blanchard and Kahn's approach to obtain the numerical solution: Let $\Gamma_1 = M\Lambda M^{-1}$, where Λ and M are the matrices of eigenvalues and eigenvectors, respectively. Without loss of generality, let us assume that $|\lambda_1|, |\lambda_2| < 1, |\lambda_3| > 1$. Then, expression (6.175) will be equivalent to the system:

$$m_{11}E_t\hat{c}_{t+1} + m_{12}\hat{k}_{t+1} + m_{13}\hat{i}_t = \lambda_1 \left(m_{11}\hat{c}_t + m_{12}\hat{k}_t + m_{13}\hat{i}_{t-1} \right) + L_1\hat{A}_t + Q_1\varepsilon_{i,t} \quad (6.176)$$

$$m_{21}E_t\hat{c}_{t+1} + m_{22}\hat{k}_{t+1} + m_{23}\hat{i}_t = \lambda_2 \left(m_{21}\hat{c}_t + m_{22}\hat{k}_t + m_{23}\hat{i}_{t-1} \right) + L_2\hat{A}_t + Q_2\varepsilon_{i,t} \quad (6.177)$$

$$m_{31}E_t\hat{c}_{t+1} + m_{32}\hat{k}_{t+1} + m_{33}\hat{i}_t = \lambda_3 \left(m_{31}\hat{c}_t + m_{32}\hat{k}_t + m_{33}\hat{i}_{t-1} \right) + L_3\hat{A}_t + Q_3\varepsilon_{i,t} \quad (6.178)$$

where $L = (L_1, L_2, L_3)' = M^{-1}\Gamma_2$, $Q = (Q_1, Q_2, Q_3)' = M^{-1}\Gamma_3$. Expression (6.178) can be solved forwards applying the law of iterated expectations. Indeed, let $z_t = m_{31}\hat{c}_t + m_{32}\hat{k}_t + m_{33}\hat{i}_{t-1}$. Then,

$$z_t = \frac{L_3}{\rho_A - \lambda_3} \hat{A}_t - \frac{Q_3}{\lambda_3} \varepsilon_{i,t},$$

that is,

$$\hat{c}_t = -\frac{m_{32}}{m_{31}} \hat{k}_t - \frac{m_{33}}{m_{31}} \hat{i}_{t-1} + \frac{L_3/m_{31}}{\rho_A - \lambda_3} \hat{A}_t - \frac{Q_3}{\lambda_3 m_{31}} \varepsilon_{i,t}, \quad (6.179)$$

which is the stability condition for the system in differences (6.175), that determines consumption as a function of the two state variables.

Plugging this stability condition in (6.176) and (6.177), we can write the two state variables as functions of their own past:

$$\begin{aligned}
& \left(m_{12} - \frac{m_{11}m_{32}}{m_{31}} \right) \hat{k}_{t+1} + \left(m_{13} - \frac{m_{11}m_{33}}{m_{31}} \right) \hat{i}_t \\
= & \lambda_1 \left(m_{12} - \frac{m_{11}m_{32}}{m_{31}} \right) \hat{k}_t + \lambda_1 \left(m_{13} - \frac{m_{11}m_{33}}{m_{31}} \right) \hat{i}_{t-1} + \quad (6.180) \\
& + \left(L_1 + \frac{L_3 m_{11}/m_{31}}{\rho_A - \lambda_3} (\lambda_1 - \rho_A) \right) \hat{A}_t + \left(Q_1 - \frac{Q_3 m_{11} \lambda_1}{\lambda_3 m_{31}} \right) \hat{Q}_t \quad (6.181)
\end{aligned}$$

$$\begin{aligned}
& \left(m_{22} - \frac{m_{21}m_{32}}{m_{31}} \right) \hat{k}_{t+1} + \left(m_{23} - \frac{m_{21}m_{33}}{m_{31}} \right) \hat{i}_t \\
= & \lambda_2 \left(m_{22} - \frac{m_{21}m_{32}}{m_{31}} \right) \hat{k}_t + \lambda_2 \left(m_{23} - \frac{m_{21}m_{33}}{m_{31}} \right) \hat{i}_{t-1} + \\
& + \left(L_2 + \frac{L_3 m_{21}/m_{31}}{\rho_A - \lambda_3} (\lambda_2 - \rho_A) \right) \hat{A}_t + \left(Q_2 - \frac{Q_3 m_{21} \lambda_2}{\lambda_3 m_{31}} \right) \hat{Q}_t \quad (6.182)
\end{aligned}$$

or, in matrix form:

$$\begin{aligned}
& \left[\begin{array}{ccc} m_{12} - \frac{m_{11}m_{32}}{m_{31}} & m_{13} - \frac{m_{11}m_{33}}{m_{31}} & - \left(L_1 + \frac{L_3 m_{11}/m_{31}}{\rho_A - \lambda_3} (\lambda_1 - \rho_A) \right) \\ m_{22} - \frac{m_{21}m_{32}}{m_{31}} & m_{23} - \frac{m_{21}m_{33}}{m_{31}} & - \left(L_2 + \frac{L_3 m_{21}/m_{31}}{\rho_A - \lambda_3} (\lambda_2 - \rho_A) \right) \\ 0 & 0 & 1 \end{array} \right] \begin{bmatrix} \hat{k}_{t+1} \\ \hat{i}_t \\ \hat{A}_t \end{bmatrix} \\
= & \left[\begin{array}{ccc} \lambda_1 \left(m_{12} - \frac{m_{11}m_{32}}{m_{31}} \right) & \lambda_1 \left(m_{13} - \frac{m_{11}m_{33}}{m_{31}} \right) & 0 \\ \lambda_2 \left(m_{22} - \frac{m_{21}m_{32}}{m_{31}} \right) & \lambda_2 \left(m_{23} - \frac{m_{21}m_{33}}{m_{31}} \right) & 0 \\ 0 & 0 & \rho_A \end{array} \right] \begin{bmatrix} \hat{k}_t \\ \hat{i}_{t-1} \\ \hat{A}_{t-1} \end{bmatrix} + \\
& \left[\begin{array}{ccc} Q_1 - \frac{Q_3 m_{11} \lambda_1}{\lambda_3 m_{31}} & 0 \\ Q_2 - \frac{Q_3 m_{21} \lambda_2}{\lambda_3 m_{31}} & 0 \\ 0 & 1 \end{array} \right] \begin{bmatrix} \varepsilon_{i,t} \\ \varepsilon_{A,t} \end{bmatrix}. \tag{6.183}
\end{aligned}$$